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Kangan Li and Yashar Mehmani

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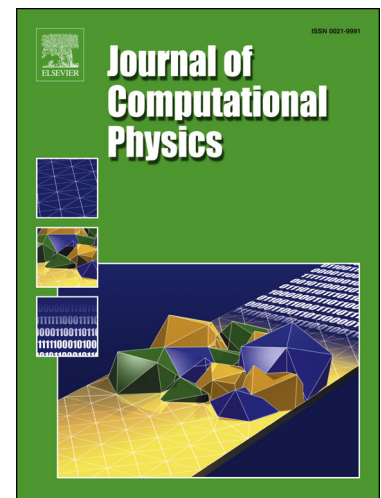
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Highlights

- Pore-level multiscale method developed for deformation of fractured porous media.
- The method decomposes a porous solid into subdomains to decouple computations.
- The decomposition need not conform to cracks, removing a key previous barrier.
- Implications for fixed-crack versus evolving-crack problems are analyzed.
- Approximation errors can be estimated and controlled iteratively.

Journal Pre-proof

A Pore-Level Multiscale Method for the Elastic Deformation of Fractured Porous Media

Kangan Li^a, Yashar Mehmani^a

^aEnergy and Mineral Engineering Department, The Pennsylvania State University, University Park, Pennsylvania 16801

Abstract

Pore-scale models are useful tools for understanding and upscaling the mechanical deformation of fractured porous media. The highest fidelity solutions are obtained by direct numerical simulation (DNS), in which governing equations are discretized and solved with a fine-grid solver (e.g., finite elements) on a 3D image (e.g., X-ray μ CT) of a porous sample. However, the downside of DNS is its high computational cost. We present a pore-level multiscale method (PLMM) that approximates DNS efficiently and with controllable accuracy in modeling the linear elastic response of porous solids with arbitrary microstructure and crack pattern. PLMM decomposes the solid into subdomains, over which local basis and correction functions are built. These functions are then coupled with a global problem to yield an approximate solution, whose errors can be iteratively corrected. A key feature of PLMM is that the decomposition need not conform to the cracks, unlike a previous variant developed by the authors. This paves the way towards solving crack nucleation and growth problems in the future without the need to dynamically update the decomposition. We represent cracks diffusely using a phase-field variable and explore three strategies for capturing them through either the basis or correction functions. The implications of each on convergence rate and computational cost are analyzed.

Keywords: Porous media, Pore-scale modeling, Multiscale method, Fracture mechanics, Phase-field method

1. Introduction

The deformation and failure of porous media is relevant to a wide range of applications from geologic CO₂ sequestration, H₂ storage, geothermal energy extraction to the design of lightweight, high-strength biomimetic materials for airplanes [1], armored vehicles [2], and thermo-acoustically insulated buildings [3]. In the subsurface, which motivates the authors, fractures are induced by either human activity (e.g., fluid injection) or tectonic stresses. In either case, they significantly impact the hydro-mechanical properties of a porous rock. To quantify these alterations, pore-scale models are useful, because they can simulate deformation and failure at the scale of individual grains and pores comprising the rock. The output from pore-scale models can be used to parameterize macroscopic models, which would otherwise contain phenomenological constants [4]. Here, we focus on the elastic deformation of fractured porous media at the pore scale, wherein cracks are assumed to be static.

With recent advances in X-ray μ CT imaging [5, 6], combined with energy-dispersive X-ray spectroscopy [7–9], the 3D microstructure and mineralogy of geologic porous samples can be captured in great detail. Such images serve as inputs to pore-scale models, which divide into two categories: direct numerical simulation (DNS) and discrete-element methods (DEM). In DNS, the governing equations are discretized on a Cartesian mesh that often coincides with the image pixels. Neither the geometry nor the equations are simplified [10]. Examples include extended finite element (XFEM) [11, 12], shifted boundary method [13, 14], immersed boundary finite volume [15, 16], phase-field models [17–19], and peri-dynamics [20, 21]. The disadvantage of DNS is that it is computationally prohibitive for analyzing large or many samples. To render computations tractable, in DEM, both the governing equations and the microstructure of a porous sample are simplified [22–24]. The solid is replaced by an assembly of “particles”

*Corresponding author: Yashar Mehmani. Email: yzm5192@psu.edu

Email addresses: kb15610@psu.edu (Kangan Li), yzm5192@psu.edu (Yashar Mehmani)

that have elementary shapes (e.g., spheres, polyhedral, clusters of spheres) [24–26], often assumed rigid. Failure is modeled by breaking fictitious “bonds” between particles when forces exceed the material’s strength [23, 27]. Recent extensions like FEM-DEM have removed the particle rigidity assumption of DEM [28, 29]. While useful in many problems, DEM and its variants have two drawbacks [30, 31]: (1) They are ill-suited for the non-particulate and highly heterogeneous porous rocks encountered in the deep subsurface; And (2) they lack a built-in mechanism to estimate and control errors. The latter is because *convergence* to a fine-scale solution is undefined.

Recently, a pore-level multiscale method (PLMM) was proposed by the authors to approximate DNS efficiently and with controllable accuracy [16] for the linear-elastic response of a fractured porous solid with arbitrary microstructure. The idea was to decompose the solid into subdomains, over which local fine-grid problems are solved to construct basis and correction functions. The decomposition was done in such a way that subdomains corresponded to physical grains in a granular medium (as in DEM), and interfaces between the subdomains to physical contacts between the grains. The basis and correction functions were then coupled to each other with a small global problem in terms of coarse-grid displacement unknowns defined at subdomain interfaces. PLMM was shown to yield an accurate initial approximation to DNS that could be corrected up to any desired accuracy using an iterative strategy. Despite these benefits, the method has three major drawbacks: (1) The decomposition must *conform* to cracks, which forbids subdomains from containing (or sustaining through later loading) any interior cracks; (2) The conformity could severely degrade the quality of the multiscale approximation in cases where cracks do not coincide with physical contacts; And (3) the iterative strategy does not correct errors near cracks, only away from them.

In this work, we reformulate PLMM [16] to remove the above limitations. The resulting algorithm approximates DNS efficiently for arbitrary cracks patterns, which need not conform to the solid’s decomposition. Moreover, PLMM errors are controllable throughout the domain. We represent cracks diffusely using a phase-field variable [32, 33], but emphasize that a sharp-crack representation is equally applicable in PLMM [11, 15]. To solve evolving-crack problems, a momentum equation must be coupled to a crack (or phase-field) evolution equation and solved repeatedly, often in a staggered fashion, at each loading step. Here, we focus only on the momentum equation, since it is more expensive to solve. However, anticipating the needs of solving evolving crack problems in the future, we propose three strategies for capturing cracks via basis and correction functions, which have important implications on convergence and computational cost: (1) All cracks are captured by the bases (Strategy A); (2) No cracks are captured by the bases (Strategy N); and (3) only some cracks, namely those that cross subdomain interfaces, are captured by the bases (Strategy U). If cracks are static, Strategy A is the fastest to converge. But if cracks are evolving, Strategy U is more efficient. This is because basis (not correction) functions are the most accurate way of capturing cracks, but their repeated update during crack evolution is costly. Strategy U strikes a delicate balance.

Multiscale methods for porous media are extensively developed at the continuum (or Darcy) scale. They include the multiscale finite element (MsFE) [12, 34–41], multiscale finite volume (MsFV) [42–47], and multiscale mortar finite element (MoMsFE) [48–51] methods to solve elliptic/parabolic equations of flow and mechanics. At the pore scale, multiscale methods are much less developed and have focused primarily on fluid flow [52–61]. The PLMM herein builds on ideas originally developed for single-phase flow [59], two-phase flow [60], and compressible gas flow [61]. A recent review of existing multiscale methods at the pore and Darcy scales can be found in [62].

The article is organized as follows: the governing equations approximated by PLMM are described in Section 2; Section 3 briefly overviews the fine-scale DNS solver employed. Section 4 discusses the algorithmic details of PLMM. Specifically, local basis and correction problems are defined in Section 4.3, the global problem that couples these local problems is formulated in Section 4.4, and the iterative strategy that corrects the PLMM errors is outlined in Section 4.5. In Sections 5–6, we validate PLMM against DNS for various 2D and 3D microstructures and a wide range of crack patterns. The results are discussed in Section 7 and the findings summarized in Section 8.

2. Problem description

Consider the porous domain $\Omega \subset \mathbb{R}^D$ with Lipschitz boundary $\partial\Omega$, where D is the number of spatial dimensions; as shown in Fig. 1a. The open set Ω is partitioned into a solid phase Ω_s and a void space Ω_v such that $\Omega = \Omega_s \cup \Omega_v \cup \Gamma^w$ and $\Omega_s \cap \Omega_v = \emptyset$, where $\Gamma^w = \overline{\Omega_s} \cap \overline{\Omega_v}$ is the interface between Ω_s and Ω_v . The black lines shown in Fig. 1a are possible cracks within Ω_s . We consider the following linear elastic momentum equations on Ω_s :

$$\nabla \cdot \sigma(\mathbf{u}) = \mathbf{f} \quad (1)$$

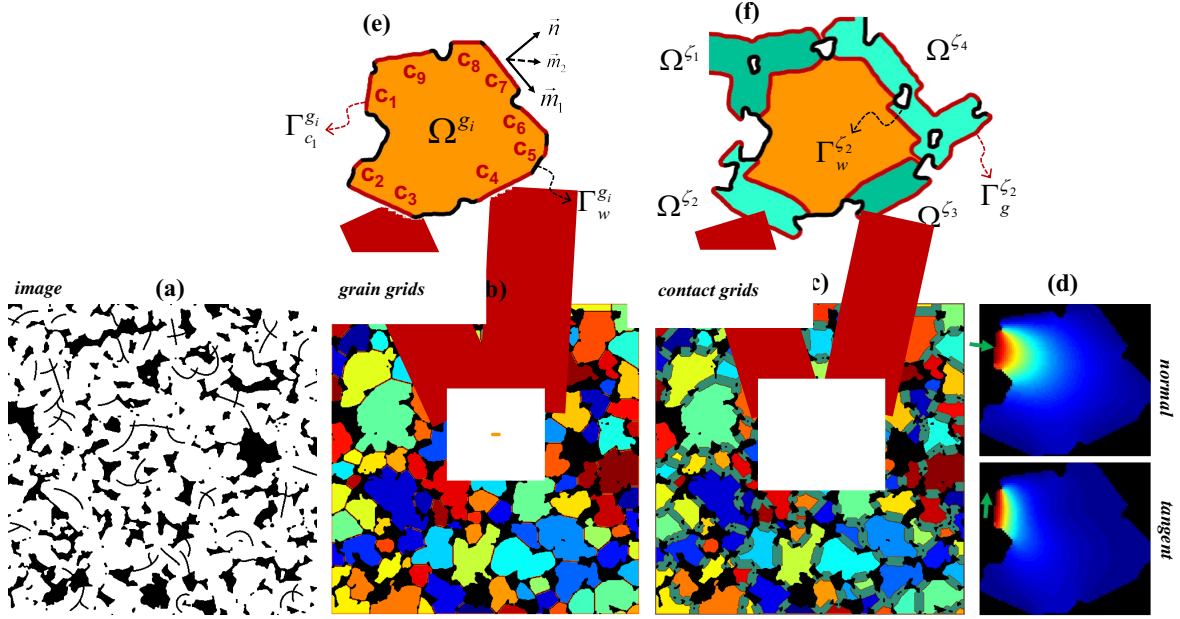


Figure 1: Schematic of a pore-scale image, its decomposition into grain grids, and contact grids. (a) The image Ω consists of a solid phase Ω_s (white), a void space Ω_v (black), and cracks Γ^F (black lines). (b) The solid Ω_s is decomposed into grain grids Ω^{g_i} (randomly colored). (c) Contact grids Ω^{ζ_k} (cyan) cover interfaces, Γ^{c_j} , between the grain grids. (d) The displacement fields of two basis functions on Ω^{g_i} are shown, corresponding to normal (compressive) and tangential (shear) BCs imposed (green arrows). (e) $\partial\Omega^{g_i}$ consists of $\Gamma_w^{g_i}$ (black) and $\Gamma_{c_1-9}^{g_i}$ (red) boundaries. The unit normal, \vec{n} , and unit tangents, $\{\vec{m}_1, \vec{m}_2\}$, on $\Gamma_{c_7}^{g_i}$ are shown. (f) The contact grids $\Omega^{\zeta_1}, \Omega^{\zeta_2}, \Omega^{\zeta_3}, \Omega^{\zeta_4}$ (shades of cyan) cover $\Gamma_{c_1, c_9}^{g_i}, \Gamma_{c_2-3}^{g_i}, \Gamma_{c_4}^{g_i}, \Gamma_{c_5-8}^{g_i}$, respectively. The boundary $\partial\Omega^{\zeta_k}$ consists of $\Gamma_w^{\zeta_k}$ (black) and $\Gamma_g^{\zeta_k}$ (red).

where \mathbf{f} is the body force, $\nabla \cdot$ the divergence operator, $\boldsymbol{\sigma}$ the stress tensor, and $\boldsymbol{\varepsilon}$ the strain tensor. The constitutive stress-strain relation for an elastic material is:

$$\boldsymbol{\sigma}(\mathbf{u}) = \mathbf{C} : \boldsymbol{\varepsilon}(\mathbf{u}) \quad (2)$$

where $\boldsymbol{\varepsilon}(\mathbf{u}) = \nabla^s \mathbf{u} = 1/2 (\nabla \mathbf{u} + \nabla^T \mathbf{u})$ and $\mathbf{C} = [C_{ijkl}]$ is the fourth-order stiffness tensor. Note ∇^s denotes the symmetric gradient operator, \mathbf{u} the displacement field, and the superscript \top transposition. For isotropic materials, \mathbf{C} equals:

$$C_{ijkl} = \lambda \delta_{ij} \delta_{kl} + \mu (\delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk}) \quad (3)$$

where λ and μ are called Lamé parameters. Substituting Eq[3] into Eq[2] we obtain:

$$\boldsymbol{\sigma}(\mathbf{u}) = \lambda \text{tr}(\boldsymbol{\varepsilon}(\mathbf{u})) \mathbf{I} + 2\mu \boldsymbol{\varepsilon}(\mathbf{u}) \quad (4)$$

in which $\text{tr}(\boldsymbol{\varepsilon})$ is the trace of $\boldsymbol{\varepsilon}$.

Let $\partial\Omega_s$ be partitioned into a Dirichlet Γ^D and a Neumann Γ^N subset, such that $\Omega_s = \overline{\Gamma^D \cup \Gamma^N}$ and $\Gamma^D \cap \Gamma^N = \emptyset$. We impose displacement boundary conditions (BCs) on Γ^D and traction BCs on Γ^N as follows:

$$\mathbf{u}|_{\Gamma^D} = \mathbf{u}_D(\mathbf{x}) \quad (5a)$$

$$\boldsymbol{\sigma}(\mathbf{u}) \cdot \mathbf{n}|_{\Gamma^N} = \mathbf{t}_N(\mathbf{x}) \quad (5b)$$

where \mathbf{u}_D is the displacement on Γ^D , and \mathbf{t}_N and \mathbf{n} are the traction and outward-pointing normal on Γ^N , respectively. With reference to Fig[1a], we assume $\Gamma^w \subset \Gamma^N$ and set $\mathbf{t}_N = \mathbf{0}$ on Γ^w , which implies the pore pressure is zero.

To represent cracks $\Gamma^F (\subset \Omega_s)$, we use a continuous damage (or phase-field) variable α satisfying [33]:

$$\alpha(\mathbf{x}) = \begin{cases} \frac{BG_c}{2l} \left(1 - \frac{\text{dist}(\mathbf{x}, \Gamma^F)}{l/2}\right), & \text{dist}(\mathbf{x}, \Gamma^F) < l/2 \\ 0, & \text{dist}(\mathbf{x}, \Gamma^F) \geq l/2 \end{cases} \quad (6)$$

where G_c is the fracture toughness, l is a length-scale parameter that controls the width of the diffuse crack, $\text{dist}(\mathbf{x}, \Gamma^F)$ is the distance from an arbitrary point \mathbf{x} to Γ^F , and B is a constant ($= 10^3$ here). If $\alpha(\mathbf{x}) = 0$, the solid is intact at point \mathbf{x} , and if $\alpha(\mathbf{x}) = 1$, it is fully damaged. In phase-field models, the variable α impacts the stress-strain Eq. 2 by degrading the stress tensor in certain directions (e.g., tension/shear but not compression) [63, 64]. Here, we choose an isotropic degradation model for simplicity [32], because our goal is to develop a multiscale approximation of DNS:

$$\boldsymbol{\sigma}(\mathbf{u}) = g(\alpha) \mathbf{C} : \boldsymbol{\varepsilon}(\mathbf{u}) \quad (7)$$

In Eq. 7, $g(\alpha) = (1 - \alpha)^2$ is the degradation function. If the solid is fully damaged, $g(\alpha) = 0$ and a traction-free condition is imposed on Γ^F . If the solid is intact, $g(\alpha) = 1$ and Eq. 7 reduces to Eq. 2. We note that more complex degradation functions also exist and can be used in PLMM without loss of generality [65]. We also note that if an anisotropic degradation model were used, PLMM would apply to the linearized form of the governing equations.

3. Single-scale modeling

The porous solid Ω_s is often represented by a digitized image (e.g., X-ray μ CT) at the pore scale, as shown by Fig. 1a. The pixels of this image can be used to construct a Cartesian *fine grid*, over which Eqs. 1 and 7 are discretized and solved with a *fine-scale solver*. The fine grid need not have the same size as the image pixels, and is often refined along the coordinate directions. We choose the finite element method as our fine-scale solver, but note that others apply equally throughout this work (e.g., finite volume [66]). We define *single-scale modeling* as the act of using the fine-scale solver over Ω_s without any approximation or domain decomposition. This yields the DNS reference solution against which PLMM is later compared.

4. Multiscale modeling

We propose a pore-level multiscale method (PLMM) to approximate, and thereby accelerate, DNS. In PLMM, the following steps are executed: (1) Ω_s is decomposed into subdomains, or coarse grids, in a way that is non-conforming to Γ^F ; (2) Basis and correction functions are built on each coarse grid by numerically solving local fine-scale problems; (3) The basis and correction functions are coupled by solving a global coarse-scale problem; And (4) a global fine-scale solution to Eqs. 1 and 7 on Ω_s is obtained by assembling the coupled basis and correction functions. These steps are detailed in Sections 4.1-4.4, which yield an initial approximation to DNS. In Section 4.5, we present an iterative approach to correct the errors of the approximation to any desired level. We explore different strategies for incorporating cracks into the basis and correction functions in Section 4.3.

4.1. Domain decomposition

To decompose Ω_s , we modify the algorithm proposed by Mehmani *et al.* [16] slightly. The approach involves applying watershed segmentation, a well-known morphological operation in image analysis [67], to Ω_s . This partitions Ω_s into N^g non-overlapping subdomains, or *grain grids*, Ω^{g_i} such that $\Omega_s = \cup_i \Omega^{g_i}$, as shown in Fig. 1b. Because of the unique properties of watershed segmentation, the subdomains correspond to local enlargements (or grains) in Ω_s , and interfaces between subdomains correspond to local constrictions (or contacts) in Ω_s . We refer to a shared interface between two adjacent grain grids as a *contact interface*, Γ^{c_j} . Notice from Fig. 1b-c that Γ^F is allowed to intersect Γ^{c_j} . In other words, the decomposition need not conform to the cracks, which is very different from, and much more flexible than, the conforming decomposition proposed in [16]. The interested reader is referred to [5, 59, 67] for further details on the watershed segmentation and domain decomposition.

We define another set of coarse grids, Ω^{ζ_k} , called *contact grids* that overlap and cover a thin neighborhood around each contact interface, Γ^{c_j} . As seen from Figs. 1c and 1f, contact grids (cyan) do not cover all of Ω_s (i.e., $\Omega_s \neq \cup_k \Omega^{\zeta_k}$).

We create Ω^{ζ_k} by morphologically “dilating”, an operation in image analysis [68], the image pixels overlaying Γ^{c_j} . The thickness of Ω^{ζ_k} can be adjusted by the user, although ~ 16 pixels has been shown to be sufficient in practice (8 per contact side) [16, 59]. More details on how Ω^{ζ_k} are constructed can be found in [59]. We refer to Ω^{g_i} and Ω^{ζ_k} as *coarse grids* and note that they are aggregates of many fine grids, which, in this work, conform at $\Omega^{g_i} \cap \Omega^{\zeta_k}$. Such conformity is not required by PLMM and different fine grids (e.g., unstructured) may, in principle, be used on Ω^{g_i} and Ω^{ζ_k} as long as appropriate mappings between the two can be defined.

4.2. Mathematical notation

In the following sections, we adopt a notation similar to [16] to formulate PLMM. Let the indices g_i, c_j, ζ_k enumerate entities associated with $\Omega^{g_i}, \Gamma^{c_j}, \Omega^{\zeta_k}$, respectively. The boundary $\partial\Omega^{g_i}$ is partitioned as $\partial\Omega^{g_i} = \Gamma_w^{g_i} \cup \Gamma_{c_j}^{g_i}$, where $\Gamma_w^{g_i} = \partial\Omega^{g_i} \cap \Gamma^w$ is the void-solid interface (black in Fig. 1e) and $\Gamma_{c_j}^{g_i} = \partial\Omega^{g_i} \cap \Gamma^{c_j}$ is a contact interface (red in Fig. 1e). Similarly, $\partial\Omega^{\zeta_k}$ consists of $\Gamma_w^{\zeta_k} = \partial\Omega^{\zeta_k} \cap \Gamma^w$ (black in Fig. 1f) and $\Gamma_{g_i}^{\zeta_k} = \partial\Omega^{\zeta_k} \cap \Omega^{g_i}, \forall g_i$ (red in Fig. 1f), where the latter corresponds to the portion of $\partial\Omega^{\zeta_k}$ that intersects the interior of grain grids (see also Fig. 2). If $\partial\Omega^{g_i}$ or $\partial\Omega^{\zeta_k}$ intersect the bounding box of $\Omega_s (= \partial\Omega_s \setminus \Gamma^w, \text{ i.e., four sides of Fig. 1a})$, then the global BCs from Eq. 5 are inherited.

Because two grain grids may share more than one contact interface (see Fig. 2), and one contact grid may cover multiple interfaces (see Figs. 1, 2), we define the following mappings between g_i, c_j and ζ_k : (1) $C^{g_i} = \{c_j | \Gamma_{c_j}^{g_i} \neq \emptyset\}$ is the index set of all Γ^{c_j} that are subsets of $\partial\Omega^{g_i}$; (2) $G^{c_j} = \{g_i | \Gamma_{c_j}^{g_i} \neq \emptyset\}$ is the index set of all grain grids sharing Γ^{c_j} . Since Γ^{c_j} is shared by exactly two grain grids, G^{c_j} has only two members ($= \{g_{i_1}, g_{i_2}\}$); (3) $Z^{c_j} = \{\zeta_k | \Gamma^{c_j} \subseteq \Omega^{\zeta_k}\}$ is the index set of all contact grids covering Γ^{c_j} . Notice Z^{c_j} has only one member, which we denote by ζ^{c_j} ; And (4) $C^{\zeta_k} = \{c_j | \Gamma^{c_j} \subseteq \Omega^{\zeta_k}\}$ is the index set of all contact interfaces covered by Ω^{ζ_k} . We use the symbol $\#$ to denote the number of members in a set. For example, $\#C^{g_i} = 9$ in Fig. 1e, $\#C^{\zeta_4} = 4$ in Fig. 1f, $\#G^{c_j} = 2$, and $\#Z^{c_j} = 1$. Finally, we use N^g, N^c, N^ζ to denote the total number of grain grids, contact interfaces, and contact grids in Ω_s .

The terms *fine-scale* and *coarse-scale* refer to entities associated with the fine grid (Section 3) and coarse grid (Ω^{g_i} or Ω^{ζ_k}), respectively. Fine-scale variables are specified by superscript f , while coarse-scale variable by superscript o . An entity is *local* if it is defined on only one coarse grid, and *global* if it is defined over the whole domain Ω_s . To simplify the exposition, we use a 2D notation to present all PLMM equations. For example, the symbol \mathbf{m} is used to denote the tangent vector on a boundary or interface (Fig. 1e). In 2D, \mathbf{m} is a single vector, but in 3D, it consists of two orthonormal vectors \mathbf{m}_1 and \mathbf{m}_2 . We use the former compact notation to express tangential BCs.

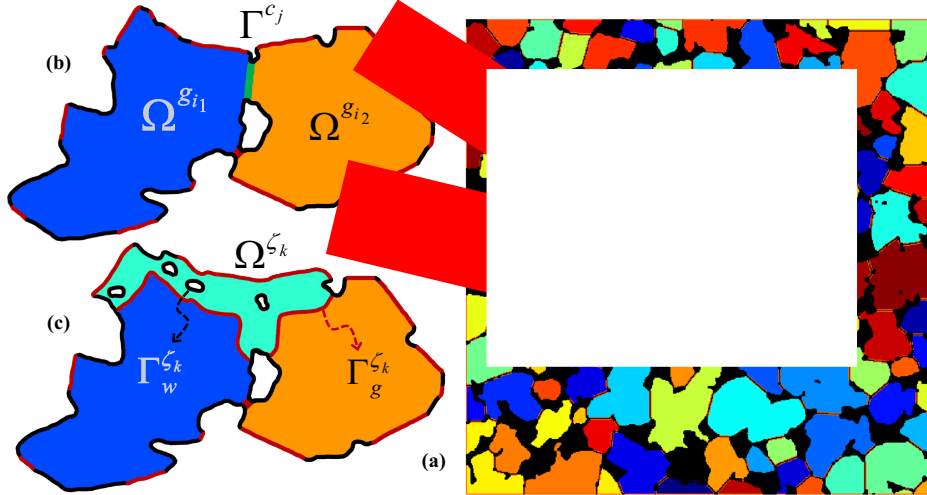


Figure 2: Schematic of a contact interface Γ^{c_j} (green) between two grain grids $\Omega^{g_{i1}}$ and $\Omega^{g_{i2}}$. Notice Γ^{c_j} is one of two shared contacts. $\Gamma_{c_j}^{g_{i1}}$ corresponds to the $-$ side of Γ^{c_j} and $\Gamma_{c_j}^{g_{i2}}$ to its $+$ side. The global problem imposes continuity of traction and displacement on Γ^{c_j} . (c) The contact grid Ω^{ζ_k} (cyan) covers Γ^{c_j} , among other contacts (red in b).

4.3. Local problems

The goal of PLMM is to solve Eqs [1] and [7] restricted to each grain grid Ω^{g_i} :

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \mathbf{u}_{g_i}^f) = \mathbf{f} \quad s.t. \quad \begin{cases} \mathbf{u}_{g_i}^f \cdot \mathbf{n} |_{\Gamma_{c_j}^{g_i}} = h_{g_i c_j 1}(\mathbf{x}), & \forall c_j \in C^{g_i} \\ \mathbf{u}_{g_i}^f \cdot \mathbf{m} |_{\Gamma_{c_j}^{g_i}} = h_{g_i c_j 2}(\mathbf{x}), & \forall c_j \in C^{g_i} \end{cases} \quad (8)$$

where $h_{g_i c_j 1}(\mathbf{x})$ and $h_{g_i c_j 2}(\mathbf{x})$ are normal and tangential displacements on $\Gamma_{c_j}^{g_i}$, respectively, and \mathbf{x} is the position vector on $\Gamma_{c_j}^{g_i}$. Given the BCs on $\Gamma_w^{g_i}$ are stress-free (i.e., $\sigma_{g_i}^f(\mathbf{u}) \cdot \mathbf{n} = \mathbf{0}$), we omit them from all subsequent local problems for brevity. Note that in 3D, the tangential displacement BC in Eq [8] consists of two separate equations corresponding to two orthonormal tangents on $\Gamma_{c_j}^{g_i}$. If Eq [8] is solved, the global fine-scale solution can be assembled from the local solutions $\mathbf{u}_{g_i}^f$ on each grain grid Ω^{g_i} . However, because $h_{g_i c_j 1}(\mathbf{x})$ and $h_{g_i c_j 2}(\mathbf{x})$ are unknown, Eq [8] cannot be solved. To make progress, we introduce the following *localization assumption*:

$$h_{g_i c_j d}(\mathbf{x}) \approx u_{g_i c_j d}^o, \quad d = 1, \dots, D \quad (9)$$

in which we have replaced the functions $h_{g_i c_j 1}(\mathbf{x})$ and $h_{g_i c_j 2}(\mathbf{x})$ with unknown scalars $u_{g_i c_j 1}^o$ and $u_{g_i c_j 2}^o$. The latter are referred to as *coarse-scale unknowns*. The localization assumption is justified by the fact that $\Gamma_{c_j}^{g_i}$ corresponds to a local constriction of Ω_s (i.e., $\Gamma_{c_j}^{g_i}$ has a small measure), which is a unique feature of the decomposition discussed in Section [4.1]. It is, therefore, reasonable to neglect the dependence of the BCs in Eq [8] on the position vector \mathbf{x} .

Substituting Eq [9] into Eq [8], we obtain:

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \mathbf{u}_{g_i}^f) = \mathbf{f} \quad s.t. \quad \begin{cases} \mathbf{u}_{g_i}^f \cdot \mathbf{n} |_{\Gamma_{c_j}^{g_i}} = u_{g_i c_j 1}^o, & \forall c_j \in C^{g_i} \\ \mathbf{u}_{g_i}^f \cdot \mathbf{m} |_{\Gamma_{c_j}^{g_i}} = u_{g_i c_j 2}^o, & \forall c_j \in C^{g_i} \end{cases} \quad (10)$$

The main advantage of Eq [10] over Eq [8] is that the solution of the former can be written as the superposition of a set of numerically-constructed *basis functions*, $\varphi_{g_i c_k d}^f$, and one *correction function*, $\tilde{\varphi}_{g_i}^f$, on Ω^{g_i} :

$$\mathbf{u}_{g_i}^f = \tilde{\varphi}_{g_i}^f + \sum_{\forall c_k \in C^{g_i}} \sum_{d=1}^D u_{g_i c_k d}^o \varphi_{g_i c_k d}^f \quad (11)$$

for arbitrary $u_{g_i c_j d}^o$. In Sections [4.3.1] [4.3.3], we present three approaches, referred to as Strategies A, N, and U, for computing $\varphi_{g_i c_k d}^f$ and $\tilde{\varphi}_{g_i}^f$. They differ in the way the solution around cracks is captured. Section [4.4] outlines how $u_{g_i c_j d}^o$ is computed. Given $\mathbf{u}_{g_i}^f$ from Eq [11], PLMM produces a global approximate fine-scale solution via:

$$\mathbf{u}^f |_{\Omega_s} = \sum_{\forall g_i} \mathbf{u}_{g_i}^f \quad (12)$$

where each $\mathbf{u}_{g_i}^f$ is extended to zero outside of Ω^{g_i} .

4.3.1. Strategy A: include crack in all basis functions

The simplest approach for computing $\varphi_{g_i c_k d}^f$ and $\tilde{\varphi}_{g_i}^f$ is to solve the following *basis problem*:

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \varphi_{g_i c_k d}^f) = \mathbf{0} \quad s.t. \quad \begin{cases} \varphi_{g_i c_k d}^f \cdot \mathbf{n} |_{\Gamma_{c_j}^{g_i}} = \delta_{kj} \delta_{1d}, & \forall c_j \in C^{g_i} \\ \varphi_{g_i c_k d}^f \cdot \mathbf{m} |_{\Gamma_{c_j}^{g_i}} = \delta_{kj} \delta_{2d}, & \forall c_j \in C^{g_i} \end{cases} \quad (13)$$

and the following *correction problem*:

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \tilde{\varphi}_{g_i}^f) = \mathbf{f} \quad s.t. \quad \begin{cases} \tilde{\varphi}_{g_i}^f \cdot \mathbf{n} |_{\Gamma_{c_j}^{g_i}} = 0, & \forall c_j \in C^{g_i} \\ \tilde{\varphi}_{g_i}^f \cdot \mathbf{m} |_{\Gamma_{c_j}^{g_i}} = 0, & \forall c_j \in C^{g_i} \end{cases} \quad (14)$$

on Ω^{g_i} . In Eq.13, δ_{kj} , δ_{1d} , and δ_{2d} are the Kronecker delta. If $d=1$, $\varphi_{g_i c_k d}^f$ corresponds to setting the normal displacement on $\Gamma_{c_k}^{g_i}$ to one, and the tangential displacement on $\Gamma_{c_k}^{g_i}$ and the normal/tangential displacements on $\Gamma_{c_j}^{g_i}$, $\forall j \neq k \in C^{g_i}$ to zero. Similarly, if $d=2$, only the tangential displacement on $\Gamma_{c_k}^{g_i}$ is set to one, while all other displacement BCs on all contacts are set to zero (Fig.1d). The basis problem accounts for inhomogeneities due to the BCs of Eq.10 and the correction problem for the inhomogeneity caused by the body force f . If parts of $\partial\Omega^{g_i}$ intersect the external boundary of Ω_s ($=\partial\Omega_s \setminus \Gamma^w$), then Eqs.13 and 14 inherit the global BCs in Eq.5. However, the right-hand side (RHS) of these BCs is set to zero for Eq.13 (homogeneous) but not Eq.14 (inhomogeneous). Given $\varphi_{g_i c_k d}^f$ and $\tilde{\varphi}_{g_i}^f$ from Eqs.13 and 14, it is easy to verify that Eq.11 satisfies Eq.10. We call this approach for computing $\varphi_{g_i c_k d}^f$ and $\tilde{\varphi}_{g_i}^f$ Strategy A, a key feature of which is that the degradation function, $g(\alpha)$, is included in the basis problems of *all* grain grids.

Remark 1. Eq.13 entails solving $\#C^{g_i} \times D$ linear systems on Ω^{g_i} , one per contact interface per normal/tangential BC, whereas Eq.14 entails solving only one system. If cracks evolve, so does $g(\alpha)$, and the basis functions must be updated at every loading step, which can become computationally expensive. To reap maximum speedup, basis functions must be updated sparingly, which makes Strategy A best suited for fixed-crack, not evolving-crack, problems. Note the basis and correction problems are fully decoupled across all grain grids and can be solved in parallel.

4.3.2. Strategy N: exclude cracks from all basis functions

Since updating basis functions in evolving-crack problems is expensive, a naïve strategy is to not update them at all. In other words, we may altogether neglect the presence of cracks in formulating the basis problem on Ω^{g_i} :

$$\nabla \cdot (\mathbf{C} : \nabla^s \varphi_{g_i c_k d}^f) = 0 \quad s.t. \quad \begin{cases} \varphi_{g_i c_k d}^f \cdot \mathbf{n}|_{\Gamma_{c_j}^{g_i}} = \delta_{kj} \delta_{1d}, & \forall c_j \in C^{g_i} \\ \varphi_{g_i c_k d}^f \cdot \mathbf{m}|_{\Gamma_{c_j}^{g_i}} = \delta_{kj} \delta_{2d}, & \forall c_j \in C^{g_i} \end{cases} \quad (15)$$

We call this Strategy N. Notice the degradation function, $g(\alpha)$, does not appear in Eq.15 and $\varphi_{g_i c_k d}^f$ corresponds to an intact grain grid. Hence, Eq.15 need only be solved once. To account for cracks, a modified correction problem is needed because if we substitute $\tilde{\varphi}_{g_i}^f$ from Eq.14 into Eq.11, we obtain $\mathbf{u}_{g_i}^f$ that does not satisfy Eq.10, unless Ω^{g_i} is intact. In other words, the following *source residual*:

$$\mathfrak{R}_s(\mathbf{u}_{g_i}^f) = \mathbf{f} - \nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \mathbf{u}_{g_i}^f) \quad (16)$$

will be non-zero. To drive \mathfrak{R}_s to zero, we propose an iterative scheme in which the modified correction problem:

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \delta \tilde{\varphi}_{g_i}^{f,\eta}) = \mathfrak{R}_s(\mathbf{u}_{g_i}^{f,\eta-1}) \quad s.t. \quad \begin{cases} \tilde{\varphi}_{g_i}^f \cdot \mathbf{n}|_{\Gamma_{c_j}^{g_i}} = 0, & \forall c_j \in C^{g_i} \\ \tilde{\varphi}_{g_i}^f \cdot \mathbf{m}|_{\Gamma_{c_j}^{g_i}} = 0, & \forall c_j \in C^{g_i} \end{cases} \quad (17)$$

is solved to obtain the *incremental correction function*, $\delta \tilde{\varphi}_{g_i}^{f,\eta}$ on Ω^{g_i} at iteration η . Notice the RHS of Eq.17 depends on the reconstructed solution $\mathbf{u}_{g_i}^{f,\eta-1}$ from the previous iteration. To compute it, we use a modified form of Eq.11:

$$\mathbf{u}_{g_i}^{f,\eta} = \tilde{\varphi}_{g_i}^{f,\eta} + \sum_{\forall c_k \in C^{g_i}} \sum_{d=1}^D \mathbf{u}_{g_i c_k d}^{o,\eta} \varphi_{g_i c_k d}^f \quad (18a)$$

$$\tilde{\varphi}_{g_i}^{f,\eta} = \gamma \delta \tilde{\varphi}_{g_i}^{f,\eta} + \tilde{\varphi}_{g_i}^{f,\eta-1} \quad (18b)$$

where γ is a relaxation parameter. We refer to iterations in η as *source iterations*, which we declare “converged” if $\|\mathfrak{R}_s(\mathbf{u}_{g_i}^{f,\eta})\|_{L_2} < 10^{-5}$. We denote the total number of source iterations performed by n_s . To start iterating, we set $\delta \tilde{\varphi}_{g_i}^{f,0}$ to zero, set $\tilde{\varphi}_{g_i}^{f,0}$ the solution of Eq.14, and compute $\mathbf{u}_{g_i}^{f,0}$ from Eq.11 with $\tilde{\varphi}_{g_i}^{f,0}$ substituting the correction function. The only remaining unknown in Eq.18 is $\mathbf{u}_{g_i c_k d}^{o,\eta}$, whose computation we detail in Section 4.4.

Remark 2. In Section 6, we show that setting $\gamma=1$ can cause source iterations to diverge if Γ^F intersects any of the contact interfaces of Ω^{g_i} . The remedy is to use a smaller $\gamma \in (0, 1]$. However, this slows convergence and the optimal value of γ is not known a priori. Therefore, Strategy N is not attractive for arbitrary crack patterns.

4.3.3. Strategy U: adaptively include cracks in some basis functions

An ideal formulation of basis and correction problems is one where basis functions are updated infrequently and $\gamma = 1$ can be set in Eq.18 without source iterations diverging. As stated in Remark 2, Strategy N diverges if $\gamma = 1$ and $\Gamma^F \cap \Gamma_{c_j}^{g_i} \neq \emptyset$ for some $c_j \in C^{g_i}$. In such cases, we empirically found that including Γ^F in the basis problem of Ω^{g_i} ensures convergence (see Section 6). Concretely, we formulate the basis problem on Ω^{g_i} as follows:

$$\begin{cases} \nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \boldsymbol{\varphi}_{g_i c_k d}^f) = \mathbf{0}, & \Gamma^F \cap \Gamma_{c_j}^{g_i} \neq \emptyset, & \exists c_j \in C^{g_i} \\ \nabla \cdot (\mathbf{C} : \nabla^s \boldsymbol{\varphi}_{g_i c_k d}^f) = \mathbf{0}, & \Gamma^F \cap \Gamma_{c_j}^{g_i} = \emptyset, & \forall c_j \in C^{g_i} \end{cases} \quad s.t. \quad \begin{cases} \boldsymbol{\varphi}_{g_i c_k d}^f \cdot \mathbf{n}|_{\Gamma_{c_j}^{g_i}} = \delta_{kj} \delta_{1d}, & \forall c_j \in C^{g_i} \\ \boldsymbol{\varphi}_{g_i c_k d}^f \cdot \mathbf{m}|_{\Gamma_{c_j}^{g_i}} = \delta_{kj} \delta_{2d}, & \forall c_j \in C^{g_i} \end{cases} \quad (19)$$

where the degradation function, $g(\alpha)$, is included if Γ^F intersects at least one of the interfaces of Ω^{g_i} . The corresponding correction problem then is Eq.14 (similar to Strategy A). If $\Gamma^F \cap \Gamma_{c_j}^{g_i} = \emptyset$ for all $c_j \in C^{g_i}$, then Eq.17 is used to compute the correction function (similar to Strategy N). Note that if Ω^{g_i} does not contain any cracks (i.e. $\Omega^{g_i} \cap \Gamma^F = \emptyset$), Eq.17 reduces to Eq.14 and converges in only one source iteration. We call this approach for computing basis and correction functions Strategy U. In evolving-crack problems, the basis functions of Ω^{g_i} are updated only if Γ^F crosses one of its interfaces with adjacent grain grids. The rationale for the update is that the localization assumption in Eq.9, thus the PLMM solution, degrades in accuracy. Having computed all basis and correction functions on Ω^{g_i} via one of Strategies A, N, and U, we proceed next to compute $\mathbf{u}_{g_i c_j d}^o$ by solving a global problem. We note that global problem is also needed as part of internal source iterations in Strategies N and U.

4.4. Global problem

Recall the coarse-scale unknown $\mathbf{u}_{g_i c_j d}^o$ corresponds to the normal displacement if $d = 1$, or the tangential displacement if $d = 2$ (or 3 in 3D) on $\Gamma_{c_j}^{g_i}$, respectively. For compactness, define $\mathbf{u}_{g_i c_j}^o = [\mathbf{u}_{g_i c_j}^o, \dots, \mathbf{u}_{g_i c_j D}^o]$. To compute $\mathbf{u}_{g_i c_j}^o$, we formulate a global problem by imposing *momentum balance* and *kinematic constraints* on all contact interfaces. Let Γ^{c_j} be shared between $\Omega^{g_{i1}}$ and $\Omega^{g_{i2}}$ with the two corresponding sides $\Gamma_{c_j}^{g_{i1}}$ and $\Gamma_{c_j}^{g_{i2}}$ (Fig.2b). The coarse-scale displacements associated with each side are $\mathbf{u}_{g_{i1} c_j}^o$ and $\mathbf{u}_{g_{i2} c_j}^o$, respectively, totalling $2 \times D$ unknowns. An equal number of equations on Γ^{c_j} are required to close the system. The first equation is a momentum (or force) balance:

$$\mathbf{t}_{g_{i1} c_j}^o = \mathbf{t}_{g_{i2} c_j}^o \quad (20)$$

where $\mathbf{t}_{g_{i1} c_j}^o$ and $\mathbf{t}_{g_{i2} c_j}^o$ are integrated tractions on $\Gamma_{c_j}^{g_{i1}}$ and $\Gamma_{c_j}^{g_{i2}}$, respectively. The second is displacement continuity:

$$\mathbf{u}_{g_{i1} c_j}^o = \mathbf{u}_{g_{i2} c_j}^o \quad (21)$$

across Γ^{c_j} , justified by the diffuse-crack representation adopted in Section 2.

To transform Eqs.20 and 21 into an algebraic system in terms of $\mathbf{u}_{g_i c_j d}^o$, we must express $\mathbf{t}_{g_{i1} c_j}^o$ and $\mathbf{t}_{g_{i2} c_j}^o$ in terms of $\mathbf{u}_{g_i c_j d}^o$. Following the approach of [11], we first derive an expression for $\mathbf{t}_{g_i c_j}^o$ defined on $\Gamma_{c_j}^{g_i}$ and drop the superscripts 1 and 2 momentarily for convenience. We begin by upscaling the basis functions on Ω^{g_i} via:

$$\boldsymbol{\varphi}_{g_i c_k d}^f = g(\alpha) \mathbf{C} : \nabla^s \boldsymbol{\varphi}_{g_i c_k d}^f \quad (22a)$$

$$\mathbf{t}_{g_i c_k d}^o = \frac{1}{|\Gamma^{c_j}|} \int_{\Gamma_{c_j}^{g_i}} \boldsymbol{\psi}_{g_i c_k d}^f \cdot \mathbf{ds} \quad (22b)$$

$$\mathbf{T}_{g_i c_k d}^o = [\mathbf{t}_{g_i c_k c_{j1} d}^o, \mathbf{t}_{g_i c_k c_{j2} d}^o, \dots]_{(\#C^{g_i} D) \times 1} \quad (22c)$$

$$\mathbf{B}_{g_i}^o = [\mathbf{T}_{g_i c_{k1} 1}^o | \dots | \mathbf{T}_{g_i c_{k1} D}^o | \mathbf{T}_{g_i c_{k2} 1}^o | \dots | \mathbf{T}_{g_i c_{k2} D}^o | \dots]_{(\#C^{g_i} D) \times (\#C^{g_i} D)} \quad (22d)$$

Eq.22a yields the fine-scale stress $\boldsymbol{\psi}_{g_i c_k d}^f$ associated with $\boldsymbol{\varphi}_{g_i c_k d}^f$. Eq.22b computes the spatial average of $\boldsymbol{\psi}_{g_i c_k d}^f$ along $\Gamma_{c_j}^{g_i}$, where $|\Gamma^{c_j}|$ is the measure of Γ^{c_j} . Eq.22c concatenates the $1 \times D$ coarse-scale tractions $\mathbf{t}_{g_i c_k c_{jl} d}^o$, $\forall c_{kl} \in C^{g_i}$, into a $(\#C^{g_i} D) \times 1$ vector $\mathbf{T}_{g_i c_k d}^o$. Finally, Eq.22d assembles $\mathbf{T}_{g_i c_k d}^o$, $\forall c_{kl} \in C^{g_i}$ into a matrix $\mathbf{B}_{g_i}^o$ with size $(\#C^{g_i} D) \times (\#C^{g_i} D)$. In effect, Eq.22 upscales the fine-scale basis functions $\boldsymbol{\varphi}_{g_i c_k d}^f$ on Ω^{g_i} into the small matrix $\mathbf{B}_{g_i}^o$.

Analogously, we upscale the correction function on Ω^{g_i} via:

$$\tilde{\psi}_{g_i}^f = g(\alpha) \mathbf{C} : \nabla^s \tilde{\varphi}_{g_i}^f \quad (23a)$$

$$\tilde{\mathbf{t}}_{g_i c_j}^o = \frac{1}{|\Gamma^{c_j}|} \int_{\Gamma_{c_j}^{g_i}} \tilde{\psi}_{g_i}^f \cdot \mathbf{ds} \quad (23b)$$

$$\tilde{\mathbf{T}}_{g_i}^o = [\tilde{\mathbf{t}}_{g_i c_{j_1}}^o, \tilde{\mathbf{t}}_{g_i c_{j_2}}^o, \dots]_{(\#C^{g_i} D) \times 1}^\top \quad (23c)$$

Eq. 23a computes the fine-scale stress $\tilde{\psi}_{g_i}^f$ associated with $\tilde{\varphi}_{g_i}^f$. Eq. 23b computes a spatial average of this stress on $\Gamma_{c_j}^{g_i}$ to yield the $1 \times D$ traction vector $\tilde{\mathbf{t}}_{g_i c_j}^o$. Finally, Eq. 23c concatenates these tractions into the $(\#C^{g_i} D) \times 1$ vector $\tilde{\mathbf{T}}_{g_i}^o$. In effect, Eq. 23 upscales the fine-scale correction function $\tilde{\varphi}_{g_i}^f$ on Ω^{g_i} into the small vector $\tilde{\mathbf{T}}_{g_i}^o$.

Remark 3. Regardless of whether Strategy A, N, or U is used to compute the basis and correction functions, the degradation function, $g(\alpha)$, must be included in the upscaling of $\varphi_{g_i c_{kd}}^f$ and $\tilde{\varphi}_{g_i}^f$ via Eqs. 22 and 23.

With $B_{g_i}^o$ and $\tilde{\mathbf{T}}_{g_i}^o$ computed for Ω^{g_i} , we can now express the coarse-scale traction $\mathbf{t}_{g_i c_j}^o$ on $\Gamma_{c_j}^{g_i}$ as a function of the coarse-scale displacements $\mathbf{u}_{g_i c_{j_l}}^o, \forall c_{j_l} \in C^{g_i}$ via Eq. 24a:

$$\mathbf{T}_{g_i}^o = B_{g_i}^o \mathbf{U}_{g_i}^o + \tilde{\mathbf{T}}_{g_i}^o \quad (24a)$$

$$\mathbf{U}_{g_i}^o = [\mathbf{u}_{g_i c_{j_1}}^o, \mathbf{u}_{g_i c_{j_2}}^o, \dots]_{(\#C^{g_i} D) \times 1}^\top \quad (24b)$$

$$\mathbf{T}_{g_i}^o = [\mathbf{t}_{g_i c_{j_1}}^o, \mathbf{t}_{g_i c_{j_2}}^o, \dots]_{(\#C^{g_i} D) \times 1}^\top \quad (24c)$$

Eq. 24a superposes the integrated tractions on $\Gamma_{c_j}^{g_i}$ computed from the basis and correction functions on Ω^{g_i} . Notice $\mathbf{U}_{g_i}^o$ is a $(\#C^{g_i} D) \times 1$ vector of the coarse-scale displacements $\mathbf{u}_{g_i c_{j_l}}^o, \forall c_{j_l} \in C^{g_i}$. Similarly, $\mathbf{T}_{g_i}^o$ is a $(\#C^{g_i} D) \times 1$ vector of the coarse-scale tractions $\mathbf{t}_{g_i c_{j_l}}^o, \forall c_{j_l} \in C^{g_i}$. Intuitively, Eq. 24a is an upscaled representation of the mechanical response of Ω^{g_i} , which relates the coarse-scale tractions on the grain grid's interfaces to the displacements imposed on them. In the literature, the matrix $B_{g_i}^o$ has been referred to as a *flux matrix* [22].

Given Eq. 24a, the momentum balance Eq. 20 on Γ can now be expressed as follows:

$$(\mathbf{B}_{g_{i_1}}^o \mathbf{U}_{g_{i_1}}^o + \tilde{\mathbf{T}}_{g_{i_1}}^o)_{c_j} (\mathbf{B}_{g_{i_2}}^o \mathbf{U}_{g_{i_2}}^o + \tilde{\mathbf{T}}_{g_{i_2}}^o)_{c_j} \quad (25)$$

where we have used the notation $\mathbf{t}_{g_i c_j}^o = (\mathbf{T}_{g_i}^o)_{c_j}$ for $i \in \{1, 2\}$. Specifically, $(\mathbf{T}_{g_i}^o)_{c_j}$ is the entry in $\mathbf{T}_{g_i}^o$ that corresponds to Γ^{c_j} . Note that $\mathbf{t}_{g_i c_j}^o$ not only depends on $\mathbf{u}_{g_i c_j}^o$, but also on the coarse-scale displacements on all other contacts of Ω^{g_i} . Concretely, Eq. 25 depends on $\mathbf{u}_{g_{i_1} c_n}^o, \forall c_n \in C^{g_{i_1}}$, and $\mathbf{u}_{g_{i_2} c_m}^o, \forall c_m \in C^{g_{i_2}}$. The *global problem* is obtained by assembling Eqs. 25 and 21, written for all Γ^{c_j} , into a single algebraic system:

$$\mathfrak{R}(\mathbf{U}) = \mathfrak{R}([\mathbf{u}_{g_i c_{jd}}^o]_{(2N^c D) \times 1}) = 0 \quad (26)$$

The residual vector $\mathfrak{R}(\cdot)$ consists of $(2N^c D) \times 1$ entries, which depend on the coarse-scale unknowns $\mathbf{u}_{g_i c_{jd}}^o$ arranged into a $(2N^c D) \times 1$ vector \mathbf{U} . Recall N^c is the total number of contacts interfaces. Since Eq. 21 imposes continuity of displacement across all Γ^{c_j} , the number of coarse-scale unknowns can be halved from $2N^c D$ to $N^c D$.

We now summarize the steps in PLMM thusfar: (a) Solve local fine-scale basis and correction problems on all Ω^{g_i} using Eqs. 13 and 14 for Strategy A, Eqs. 15 and 17 for Strategy N, or Eqs. 19 and 17 for Strategy U, respectively; (b) Solve the global problem, Eq. 26, to obtain coarse-scale displacements $\mathbf{u}_{g_i c_{jd}}^o$ on all Γ^{c_j} ; (c) Reconstruct the fine-scale solution $\mathbf{u}_{g_i}^f$ on each Ω^{g_i} via Eq. 11 for Strategy A, and Eq. 18 for Strategies N and U; And (d) assemble the global fine-scale solution \mathbf{u}^f on Ω_s via Eq. 12. In Strategies N and U, source iterations between Steps (a) and (c) are required to compute $\tilde{\varphi}_{g_i}^f$, and before proceeding to Step (d). The above steps for PLMM are detailed in Algorithm 1 and yield an initial approximation to single-scale DNS, which we denote by \mathbf{A}_0 , \mathbf{N}_0 , or \mathbf{U}_0 depending on the strategy used to compute basis and correction functions. We next discuss how to iteratively reduce the errors of this approximation.

Algorithm 1: PLMM without boundary iterations (output: A_0 , N_0 , or U_0)

Decompose Ω_s into non-overlapping grain grids Ω^{g_i}
Do $g_i \in \{1, \dots, N^g\}$
 Do $c_j \in C^{g_i}$
 Do $d \in \{1, \dots, D\}$
 If (Strategy A), Solve Eq.13 to obtain basis functions $\varphi_{g_i c_j d}^f$, **End if**
 If (Strategy N), Solve Eq.15 to obtain basis functions $\varphi_{g_i c_j d}^f$, **End if**
 If (Strategy U), Solve Eq.19 to obtain basis functions $\varphi_{g_i c_j d}^f$, **End if**
 End do
 End do
 Solve Eq.14 to obtain correction function $\tilde{\varphi}_{g_i}^f (= \tilde{\varphi}_{g_i}^{f,0}$ in source iterations)
 Upscale basis and correction functions via Eqs.22,23
End do
Solve global problem in Eq.26 to obtain $u_{g_i c_j d}^o, \forall g_i, c_j, d$
Reconstruct the local fine-scale solution $u_{g_i}^f$ via Eq.11
Compute initial source residual \mathfrak{R}_s^0 via Eq.16
While ($\eta < n_s$ and $\mathfrak{R}_s^\eta > tol_s$) \rightarrow source iterations
 Solve Eq.17 to obtain incremental correction function $\delta \tilde{\varphi}_{g_i}^{f,\eta}$
 Update correction function $\tilde{\varphi}_{g_i}^{f,\eta}$ via Eq.18b (set $\gamma = 1$ for Strategy A or U)
 Solve modified global problem in Eq.26 to obtain $u_{g_i c_j d}^{o,\eta}, \forall g_i, c_j, d$
 Reconstruct local fine-scale solution $u_{g_i}^{f,\eta}$ via Eq.18a
 Update the source residual \mathfrak{R}_s^η via Eq.16
End while
Assemble global fine-scale solution $u^{f,0}$ via Eq.12

4.5. Iterative correction

Errors in PLMM, measured against the single-scale solution, are due to the localization assumption, $h_{g_i c_j d}(\mathbf{x}) \approx u_{g_i c_j d}^o$ in Eq.10. To correct them, Eq.14 must be modified as follows:

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \tilde{\varphi}_{g_i}^f) = f \quad s.t. \quad \begin{cases} \tilde{\varphi}_{g_i}^f \cdot \mathbf{n}|_{\Gamma_{c_j}^{g_i}} = h_{g_i c_j 1}(\mathbf{x}) - u_{g_i c_j 1}^o, & \forall c_j \in C^{g_i} \\ \tilde{\varphi}_{g_i}^f \cdot \mathbf{m}|_{\Gamma_{c_j}^{g_i}} = h_{g_i c_j 2}(\mathbf{x}) - u_{g_i c_j 2}^o, & \forall c_j \in C^{g_i} \end{cases} \quad (27)$$

where the BCs include information that is missing from Eq.14. If Eq.27 is solved instead of Eq.14, all in PLMM errors would be removed and $u_{g_i}^f$ obtained from Eq.11 would satisfy Eq.8 exactly. However, Eq.27 cannot be solved because $h_{g_i c_j 1}(\mathbf{x})$ and $h_{g_i c_j 2}(\mathbf{x})$ are not known a priori. We therefore proceed to approximate them instead by adopting an iterative scheme, similar to ideas in [16, 45, 59]. We refer to such corrective iterations as *boundary* iterations, to differentiate them from the *source* iterations discussed in Section 4.3.

Remark 4. In Strategy A, replacing Eq.14 with Eq.27 would immediately yield an exact correction function. But in Strategies N and U, the replacement would merely provide a more accurate initial guess to source iterations (Eq.18), which ultimately converge to the exact correction function. In other words, Eq.17 would remain unaltered, implying the BCs of Eq.27 may fixed across source iterations. Because Eq.27 is unsolvable in practice, these statements apply equally to the approximated form of Eq.27 presented below (i.e., Eq.29).

Let ω be the boundary iteration index. Starting from $\omega - 1$, at which the global fine-scale solution $u^{f,\omega-1}$ is known, we proceed to compute $u^{f,\omega}$. The initial guess $u^{f,0}$ is taken to be the initial approximation produced by PLMM (Algorithm 1). To estimate the BCs of Eq.27, we first solve a *contact problem* on each contact grid Ω^{ζ_k} (Fig.1c):

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s u_{\zeta_k}^{f,\omega}) = f \quad s.t. \quad \begin{cases} u_{\zeta_k}^{f,\omega} \cdot \mathbf{n}|_{\Gamma_g^{\zeta_k}} = R_{\Gamma_g^{\zeta_k}} [u^{f,\omega-1}] \cdot \mathbf{n} \\ u_{\zeta_k}^{f,\omega} \cdot \mathbf{m}|_{\Gamma_g^{\zeta_k}} = R_{\Gamma_g^{\zeta_k}} [u^{f,\omega-1}] \cdot \mathbf{m} \end{cases} \quad (28)$$

where the operator $R_{\Gamma_{\zeta_k}^{f,\omega-1}}$ restricts $\mathbf{u}^{f,\omega-1}$ onto $\Gamma_{\zeta_k}^{f,\omega-1}$. The BCs on $\Gamma_{\zeta_k}^{f,\omega-1}$ are stress-free (black in Fig. 2c), while the Dirichlet BCs on $\Gamma_g^{f,\omega-1}$ (red in Fig. 2c) are inherited from the fine-scale solution $\mathbf{u}^{f,\omega-1}$ at the previous iteration. If $\partial\Omega^{\zeta_k}$ intersects the external boundary of Ω_s , that is $\partial\Omega^{\zeta_k} \cap (\partial\Omega_s \setminus \Gamma^w) \neq \emptyset$, then the global BCs from Eq. 5 are inherited.

After $\mathbf{u}_{\zeta_k}^{f,\omega}$, $\forall \zeta_k$ are computed, the correction problem Eq. 27 is replaced by its approximated form:

$$\nabla \cdot (g(\alpha) \mathbf{C} : \nabla^s \tilde{\boldsymbol{\varphi}}_{g_i}^{f,\omega}) = \mathbf{f} \quad \text{s.t.} \quad \begin{cases} \tilde{\boldsymbol{\varphi}}_{g_i}^{f,\omega} \cdot \mathbf{n} |_{\Gamma_{\zeta_j}^{g_i}} = \left(R_{\Gamma_{\zeta_j}^{g_i}} [\mathbf{u}_{\zeta_j}^{f,\omega}] - \langle R_{\Gamma_{\zeta_j}^{g_i}} [\mathbf{u}_{\zeta_j}^{f,\omega}] \rangle \right) \cdot \mathbf{n} \\ \tilde{\boldsymbol{\varphi}}_{g_i}^{f,\omega} \cdot \mathbf{m} |_{\Gamma_{\zeta_j}^{g_i}} = \left(R_{\Gamma_{\zeta_j}^{g_i}} [\mathbf{u}_{\zeta_j}^{f,\omega}] - \langle R_{\Gamma_{\zeta_j}^{g_i}} [\mathbf{u}_{\zeta_j}^{f,\omega}] \rangle \right) \cdot \mathbf{m} \end{cases} \quad (29)$$

where $R_{\Gamma_{\zeta_j}^{g_i}} [\cdot]$ restricts $\mathbf{u}_{\zeta_j}^{f,\omega}$ onto $\Gamma_{\zeta_j}^{g_i}$. Recall ζ_j is the index of the contact grid covering $\Gamma_{\zeta_j}^{g_i}$ ($\subset \Gamma_{\zeta_j}^{c_j}$). The operator $\langle \cdot \rangle$ spatially averages its argument over $\Gamma_{\zeta_j}^{g_i}$. Notice $R_{\Gamma_{\zeta_j}^{g_i}} [\mathbf{u}_{\zeta_j}^{f,\omega}] \cdot \mathbf{n}$ and $\langle R_{\Gamma_{\zeta_j}^{g_i}} [\mathbf{u}_{\zeta_j}^{f,\omega}] \rangle \cdot \mathbf{n}$ approximate $h_{g_i c_j 1}(\mathbf{x})$ and $u_{g_i c_j 1}^o$ in Eq. 27 respectively. The same holds for the tangential BCs in Eq. 29. Once $\tilde{\boldsymbol{\varphi}}_{g_i}^{f,\omega}$ is obtained from Eq. 29, we can compute $\mathbf{u}^{f,\omega}$ using equations similar to Eqs. 22-24. For brevity, we only list those that are updated by ω :

$$\tilde{\mathbf{t}}_{g_i c_j}^{o,\omega} = \frac{1}{|\Gamma_{\zeta_j}^{g_i}|} \int_{\Gamma_{\zeta_j}^{g_i}} \tilde{\boldsymbol{\psi}}_{g_i}^{f,\omega} \cdot d\mathbf{s} \quad (30a)$$

$$\tilde{\mathbf{T}}_{g_i}^{o,\omega} = [\tilde{\mathbf{t}}_{g_i c_{j_1}}^{o,\omega}, \tilde{\mathbf{t}}_{g_i c_{j_2}}^{o,\omega}, \dots]^T_{(\#C^{g_i} D) \times 1} \quad (30b)$$

$$\mathbf{T}_{g_i}^{o,\omega} = \mathbf{B}_{g_i}^o \mathbf{U}_{g_i}^{o,\omega} + \tilde{\mathbf{T}}_{g_i}^{o,\omega} \quad (30c)$$

$$\Re(\mathbf{U}^\omega) = \Re([\mathbf{u}_{g_i c_j d}^{o,\omega}]_{(2N^c D) \times 1}) = 0 \quad (30d)$$

$$\mathbf{u}^{f,\omega} |_{\Omega_s} = \sum_{g_i} (\tilde{\boldsymbol{\varphi}}_{g_i}^{f,\omega} + \sum_{c_k \in C^{g_i}} \sum_d u_{g_i c_k d}^{o,\omega} \boldsymbol{\varphi}_{g_i c_k d}^f) \quad (30e)$$

Notice $\mathbf{B}_{g_i}^o$ and $\boldsymbol{\varphi}_{g_i c_k d}^f$ associated with the basis functions, remain unaltered by ω . The global fine-scale solution $\mathbf{u}^{f,\omega}$ obtained from Eq. 30e converges to the single-scale solution as $\omega \rightarrow \infty$. Eqs. 28 and 29 are fully decoupled across all contact grids and grain grids and can be easily parallelized. Algorithm 2 summarizes the overall workflow of PLMM, in which n_b is the total number of boundary iterations performed. We denote the global fine-scale solution $\mathbf{u}^{f,\omega}$ obtained from Algorithm 2 by \mathbf{A}_ω , \mathbf{N}_ω , or \mathbf{U}_ω depending on the strategy used to compute basis/correction functions.

Algorithm 2: PLMM with boundary iterations (output: \mathbf{A}_ω , \mathbf{N}_ω , or \mathbf{U}_ω where $\omega > 0$)

Compute initial approximation $\mathbf{u}^{f,0}$ via Algorithm 1 (i.e., \mathbf{A}_0 , \mathbf{N}_0 , or \mathbf{U}_0)

Do $\omega \in \{1, \dots, n_b\}$

Do $\zeta_k \in \{1, \dots, N^\zeta\}$

 Solve Eq. 28 to obtain contact-grid solution $\mathbf{u}_{\zeta_k}^{f,\omega}$ (inherit BCs from $\mathbf{u}^{f,\omega-1}$)

End do

Do $g_i \in \{1, \dots, N^g\}$

 Solve Eq. 29 to obtain correction function $\tilde{\boldsymbol{\varphi}}_{g_i}^{f,\omega}$ (inherit BCs from $\bigcup_k \mathbf{u}_{\zeta_k}^{f,\omega}$)

 [Note: requires source iterations for Strategies N and U; see Algorithm 1]

End do

 Solve global problem in Eq. 30d to obtain $u_{g_i c_j d}^{o,\omega}, \forall g_i, c_j, d$

 Reconstruct and assemble global fine-scale solution $\mathbf{u}^{f,\omega}$ via Eq. 30e

End do

5. Validation set

We compare the proposed PLMM against single-scale DNS for different porous geometries shown in Fig. 3. They include a 2D disk pack (P2D) [16], a 2D sandstone (S2D) [69], and a 3D sandstone (S3D) [69]. All have Lamé parameters $\lambda = 8.3$ GPa and $\mu = 44.3$ GPa corresponding to α -quartz [16, 70]. In both P2D and S2D, the bottom

boundary of the domain ($y = 0$) is fixed and a vertical-upward displacement of $u_D = 0.004$ mm is applied to the top boundary ($y = L_y$). In S3D, the domain is clamped on the right-back surface ($x = 0$) and a displacement of $u_D = 0.008$ mm along the positive x -direction is imposed on the left-front surface ($x = L_x$). All other boundaries are stress-free.

The fine grids in Fig.3 coincide with the image pixels, which are 716×576 for P2D, 541×546 for S2D, and $134 \times 136 \times 142$ for S3D. The phase-field length-scale parameter is set to $l = 2h = 0.02$ mm, where h is the fine-grid size, and the fracture toughness to $G_c = 2.7 \times 10^{-3}$ kN/mm. Fig.3 also depicts the grain grids and contact grids of each domain obtained from the decomposition algorithm of Section 4.1. The number of grain grids, N^g , and contact grids, N^c , are 76 and 127 for P2D, 121 and 145 for S2D and 43 and 52 for S3D, respectively. Contact grids are 16 pixels wide in P2D and S2D, and 12 pixels wide in S3D.

Figs.4 and 5 depict the crack patterns considered in each domain. For P2D and S2D, they include: Pattern 1, in which all cracks are confined to the interior of the grain grids, i.e., none crosses a contact interface; Pattern 2, in which all cracks cross contact interfaces, i.e., none is confined solely to the interior of a grain grid; and Pattern 3, which is a superposition of cracks in patterns 1 and 2. We use P2D-c₁ and S2D-c₁ to refer to pattern 1, P2D-c₂ and S2D-c₂ to refer to pattern 2, and P2D-c₃ and S2D-c₃ to refer to pattern 3. For S3D, we only consider one crack pattern, which consists of 20 elliptical cracks with different sizes and orientations placed randomly in the domain. Most cracks intersect contact interfaces. This pattern is denoted by S3D-c₂₀ and is most comparable to P2D-c₃ and S2D-c₃.

Crack pattern 2 in Fig.4 consists of two subgroups: (1) cracks parallel to contact interfaces, which sever adjacent subdomains; and (2) cracks that strike interfaces at some angle. The latter presents a more significant challenge to PLMM as the displacement *along* each interface is discontinuous, invalidating the localization assumption in Eq.9. Cracks in subgroup 1 may still preserve continuity. Hence, Fig.4 considers more cracks from subgroup 2.

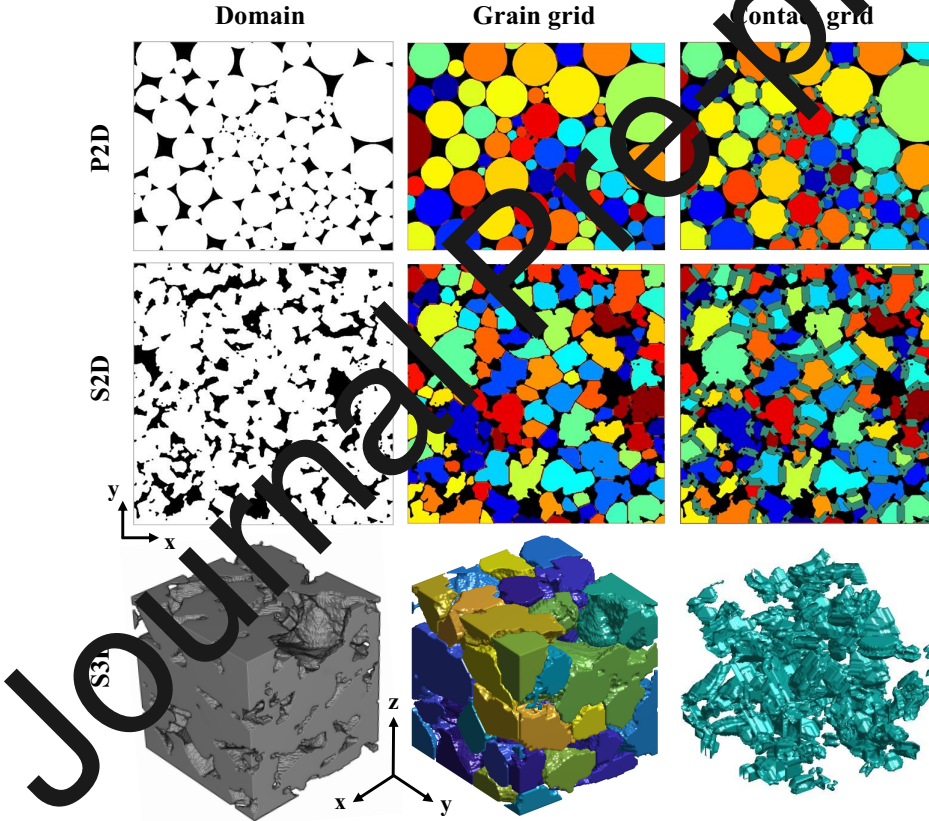


Figure 3: Domains used to validate PLMM. From top to bottom: 2D disk pack (P2D), 2D sandstone (S2D), and 3D sandstone (S3D). From left to right: domain image, grain grids (randomly colored), and contact grids (cyan).

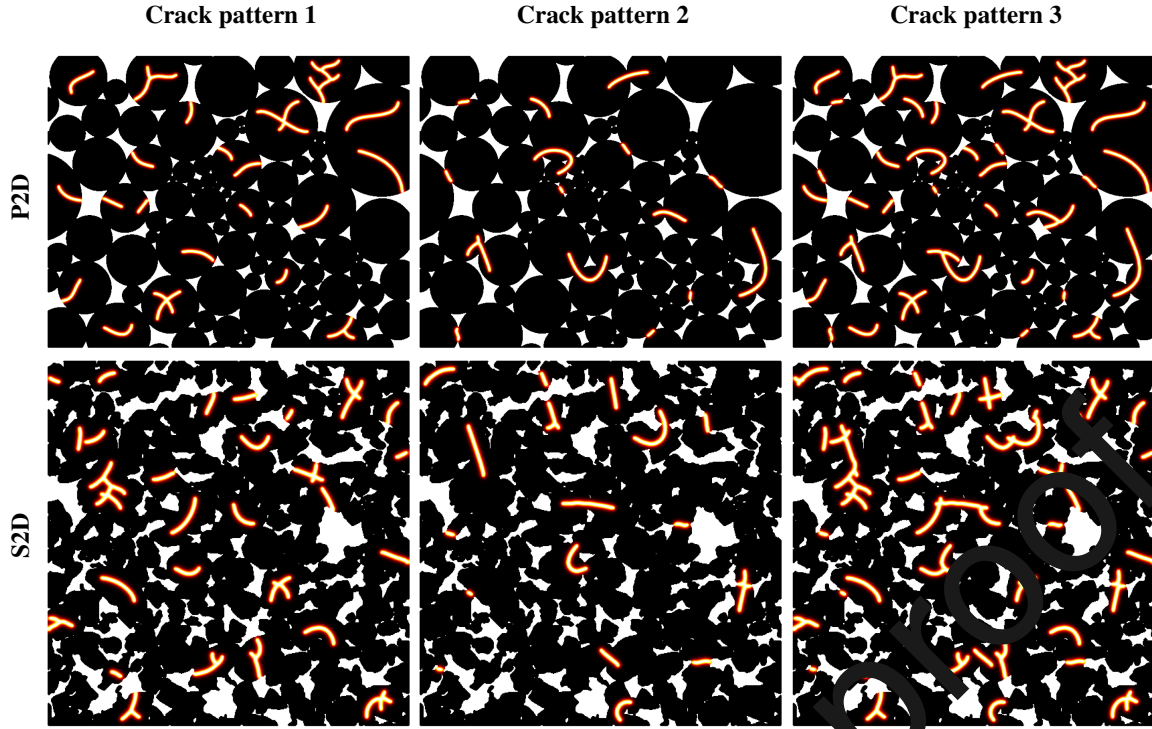


Figure 4: Crack patterns considered for the P2D and S2D domains in Fig. 3. Pattern 1: all cracks are confined to the interior of grain grids, i.e., none crosses a contact interface. Pattern 2: all cracks cross contact interfaces, i.e., none is solely confined to the interior of a grain grid. Pattern 3: superposition of cracks from patterns 1 and 2.

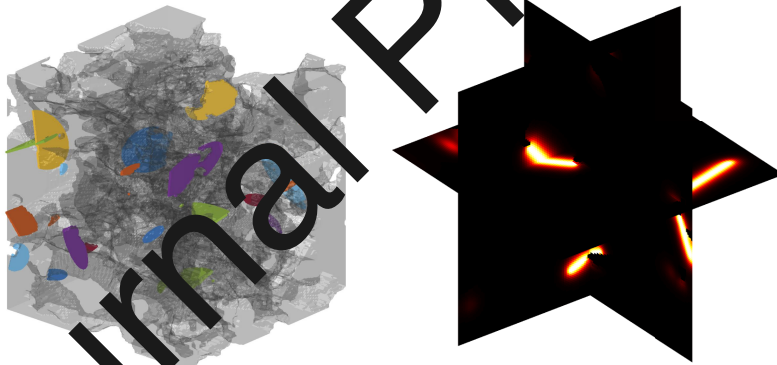


Figure 5: Crack pattern considered for the S3D domain in Fig. 3. (Left) Twenty elliptical cracks with different sizes and orientations placed randomly inside the domain. (Right) A slice plot of the phase-field variable corresponding to the cracks. The length scale parameter is set to $l=0.02$ mm.

6. Results

We solve Eqs. [1] and [7] on the pore-scale domains in Fig. [3] via PLMM and compare the solutions against single-scale DNS. To quantify the PLMM errors, we define the following error metrics similar to [16, 59]:

$$E_p^\chi = \frac{\|\chi_M - \chi_S\|}{\sup_{\Omega_s} \|\chi_S\|} \times 100, \quad E_2^\chi = \left(\frac{1}{|\Omega_s|} \int_{\Omega_s} (E_p^\chi)^2 d\Omega \right)^{1/2} \quad (31)$$

In Eq. [31] χ is a placeholder for either displacement \mathbf{u} , volumetric strain $\nabla \cdot \mathbf{u}$, or maximum shear stress σ_t . The subscripts M and S refer to variables associated with PLMM and DNS, respectively. In Eq. [31] E_p^χ measures the relative pointwise error between PLMM and DNS, and E_2^χ the relative L_2 -error, both expressed as percentages. In the following sections, PLMM solutions are computed using Strategies A, N, and U (Section [4.3]) for obtaining the basis and correction functions, and different numbers of source and boundary iterations are considered (Section [4]).

Before discussing the results, we note the PLMM errors (E_p^χ and E_2^χ) herein are the result of three contributions:

$$E = E_b + E_s + E_f \quad (32)$$

where we have dropped the superscript χ and subscripts p and 2 for clarity. E_b represents errors that are corrected by boundary iterations, E_s errors corrected by source iterations, and E_f errors reduced by only refining the fine grid. Specifically, holding the fine-grid size, h , fixed, if the number of boundary, n_b , and source, n_s , iterations $\rightarrow \infty$, then E_b and $E_s \rightarrow 0$, respectively. The remaining error, E_f , is due to fine-scale discretization differences between PLMM and DNS near contact interfaces. This is caused by our specific implementation of PLMM and can, in principle, be avoided [16, 59]. In Appendix A, we present a mesh convergence analysis that shows $E_f \sim O(h)$ for displacement.

6.1. Initial multiscale approximation vs. single-scale solution

Figs. [6]–[7] compare the initial PLMM approximation U_0 to single-scale DNS for P2D-c₃ and S2D-c₃. Results for N_0 and A_0 are similar and thus omitted. Since crack pattern 3 is more challenging than patterns 1 and 2, the latter results are omitted too. Recall that U_0 implies no boundary iterations are performed ($n_b = 0$). Here, we set the number of source iterations, n_s , to be sufficiently large to ensure convergence ($tol_s = 10^{-12}$). Figs. [6]–[7] compare the displacement magnitude $|\mathbf{u}|$, volumetric strain $|\nabla \cdot \mathbf{u}|$, and maximum shear stress σ_t from PLMM and DNS. In all cases, the agreement is very good. Fig. [8] shows a similarly good agreement between U_0 and DNS for the S3D-c₂₀ domain.

6.2. Multiscale errors vs. boundary iterations

We next quantify the PLMM errors as a function of boundary iterations. Similar to Section [6.1] we set n_s to be sufficiently large to ensure source iteration convergence ($tol_s = 10^{-12}$). We continue to focus on Strategy U because the results for Strategies A and N are similar. Fig. [9] depicts pointwise errors for U_0 , U_4 , and U_8 in P2D-c₃. The errors correspond to the displacement, \mathbf{u} , volumetric strain, $\nabla \cdot \mathbf{u}$, and maximum shear stress, σ_t . Focusing on U_0 , we see that $E_p^u < 8\%$, $E_p^{\nabla \cdot u} < 1\%$, and $E_p^{\sigma_t} < 1\%$ in most parts of the domain except near cracks and contact interfaces. The errors near contacts are due to the localization assumption and are most obvious for $\nabla \cdot \mathbf{u}$ and σ_t . We note that the U_0 errors correspond to the PLMM solution depicted in Fig. [6]. Focusing next on U_4 and U_8 , we see that after 4 and 8 boundary iterations, respectively, errors decrease throughout the domain. The relative L_2 -error for displacement, E_2^u , is 3.28% for U_0 , 2.28% for U_4 , and 2.00% for U_8 . The apparent stagnation in errors after U_4 is due to the fine-grid discretization error E_f in Eq. [32] which is $O(h)$ and can be reduced only by refining the fine grid (see Appendix A).

Almost identical observations are made for S2D-c₃ and S3D-c₂₀ based on pointwise errors shown in Figs. [10] and [11], respectively. Focusing on S2D-c₃, the U_0 errors are $E_p^u < 10\%$, $E_p^{\nabla \cdot u} < 0.8\%$, and $E_p^{\sigma_t} < 10\%$ in most parts of the domain except near cracks and contact interfaces. However, after 4 and 8 boundary iterations, errors decrease throughout the domain and E_2^u drops from 8.08% for U_0 , to 4.87% for U_4 , to 3.60% for U_8 . The same trend is seen for S3D-c₂₀, where E_2^u decreases from 2.80% for U_0 , to 1.66% for U_4 , to 1.57% for U_8 .

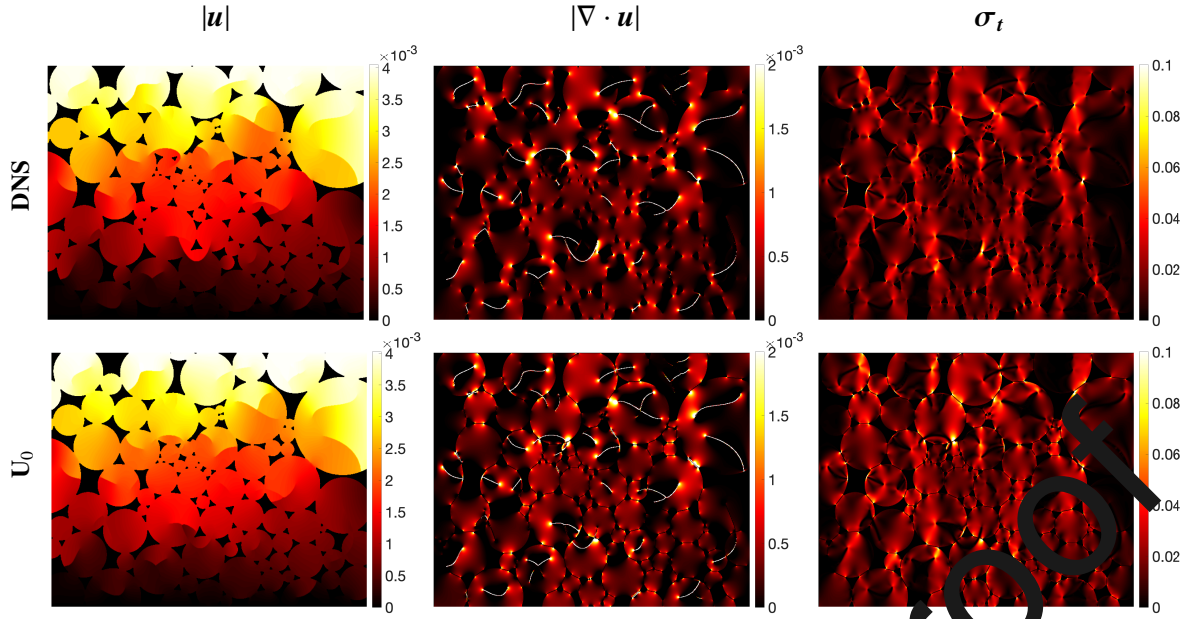


Figure 6: Comparison between the multiscale approximation U_0 and single-scale DNS for P2D- c_3 . From left to right, the displacement magnitudes $|u|$, volumetric strains $|\nabla \cdot u|$, and maximum shear stresses σ_t are shown.

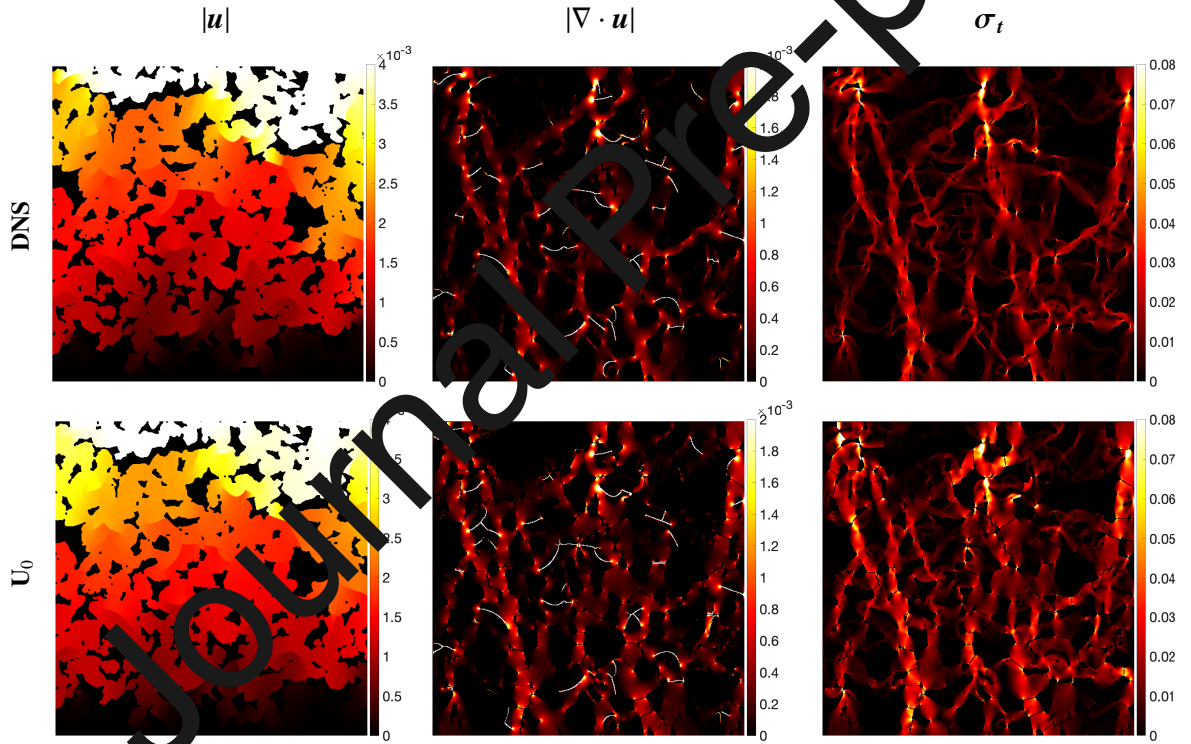


Figure 7: Comparison between the multiscale approximation U_0 and single-scale DNS for S2D- c_3 . From left to right, the displacement magnitudes $|u|$, volumetric strains $|\nabla \cdot u|$, and maximum shear stresses σ_t are shown.

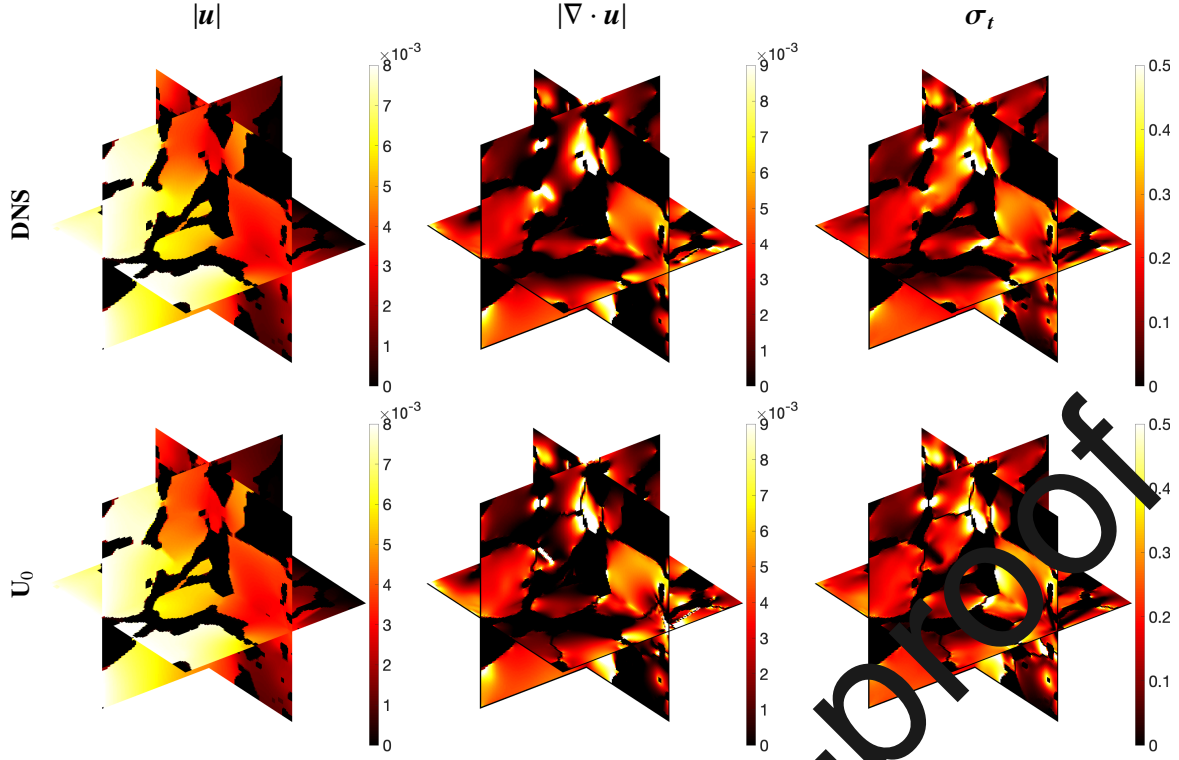


Figure 8: Comparison between the multiscale approximation U_0 and single-scale DNS for S3D-c₂₀. From left to right, the displacement magnitudes $|u|$, volumetric strains $|\nabla \cdot u|$, and maximum shear stresses σ_t are shown.

6.3. Multiscale errors vs. source iterations

In Section 6.2 we discussed PLMM errors for when n_s is sufficiently large for source iterations to converge at *each* boundary iteration. This is computationally wasteful, as smaller n_s can still ensure convergence to the DNS solution. Here, we analyze Strategies A, N, and U for different combinations of n_s and n_b to understand their convergence rates. We focus only on S2D and P2D with crack patterns 1 and 2 due to the large number of simulations required.

Fig 12 depicts contour plots of E_2^u versus n_s and n_b for Strategies A, N, and U applied to P2D-c₁ and P2D-c₂. The following key observations can be made: (1) Regardless of the crack pattern, Strategy A does not require any source iterations to converge, as expected from Section 4.3.1; (2) When all cracks are confined to the interior of grain grids (pattern 1), Strategy U reduces to Strategy N, consistent with Eqs. 19 and 15; (3) When all cracks intersect contact interfaces, Strategy U reduces to Strategy A, consistent with Eqs. 19 and 13; (4) In Strategies N and U, source iterations need not fully converge for PLMM to converge to the DNS solution; And (5) convergence to the DNS solution is slower in all strategies when cracks intersect contact interfaces (pattern 2), because the localization assumption (Eq. 9) becomes degraded. Similar observations can be made for S2D-c₁ and S2D-c₂ from the contour plots of E_2^u shown in Fig. 13. We note that Strategy A converges the fastest, in ~ 8 boundary iterations, regardless of the crack pattern, which is comparable to the results reported in [16] for intact domains.

In Figs. 12 and 13, we ensure that source iterations in Strategy N converge in crack pattern 2, a relaxation parameter of $\gamma = 0.3$ was used in Eq. 18b. In fact, larger values of γ caused Strategy N to *diverge*, as shown by Fig. 14 for $\gamma = 0.6$ and 0.9 . We note that relaxing source iterations was deemed necessary only in Strategy N and in the presence of cracks intersecting contact interfaces (patterns 2 and 3). In all other cases, $\gamma = 1$ was found sufficient and used. This renders Strategies A and U, which do not require relaxation, superior to Strategy N, because not only does $\gamma < 1$ slow down convergence, but the exact value of γ cannot be determined *a priori* (at least based on this work).

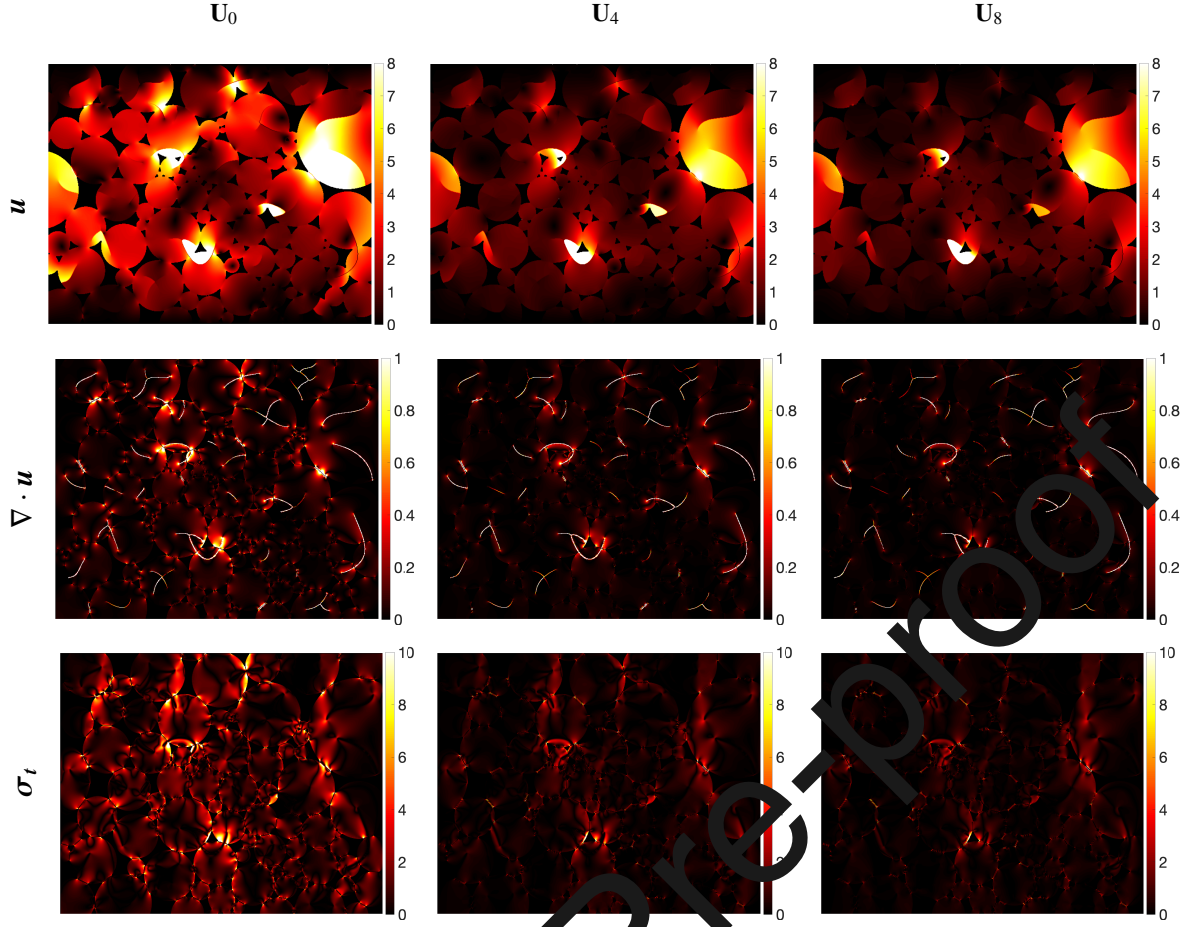


Figure 9: Pointwise errors (%) in displacement (E_p^u), volumetric strain ($E_p^{\nabla \cdot u}$), and maximum shear stress ($E_p^{\sigma_t}$) for the PLMM solutions U_0 , U_4 , and U_8 in the domain P2D-c₃. The U_0 plots correspond to Fig. 6.

7. Discussion

7.1. Convergence rate for different strategies

Figs. 12-14 show that Strategy A converges the fastest, followed closely by Strategy U. Strategy N is the slowest to converge, and may even diverge if not relaxed in crack patterns 2 and 3. The reason for divergence is that when cracks intersect contact interfaces, the localization assumption (Eq. 9) becomes degraded and boundary iterations are required to correct it. In addition, since cracks are excluded from the basis functions of Strategy N, source iterations are needed to capture displacement discontinuities within grain grids. Hence, the correction problem (Eq. 17) attempts to correct too much, leading to an overshoot in the correction increment $\delta \phi_{g_i}^{f,\eta}$. Relaxation tempers this increment, thereby encouraging convergence. Strategies A and U, by contrast, include contact-intersecting cracks in their basis functions, requiring only boundary iterations to correct for inaccuracies in the localization assumption. Therefore, the correction problem has less to correct, obviating relaxation. If cracks are fully confined to the interior of grain grids (pattern 1), the localization assumption is not degraded, and Strategies N and U require only source iterations to yield an accurate approximate solution; because neither strategy includes interior-cracks in their basis problems. Strategy A also includes interior cracks in its bases, and thus, does not even require source iterations.

7.2. Fixed-crack vs. evolving-crack problems

Given Strategy A converges the fastest, then why bother with Strategies N and U? If cracks are fixed, the answer is: we should not. Strategy A is the most efficient computationally. But in problems where cracks evolve over many

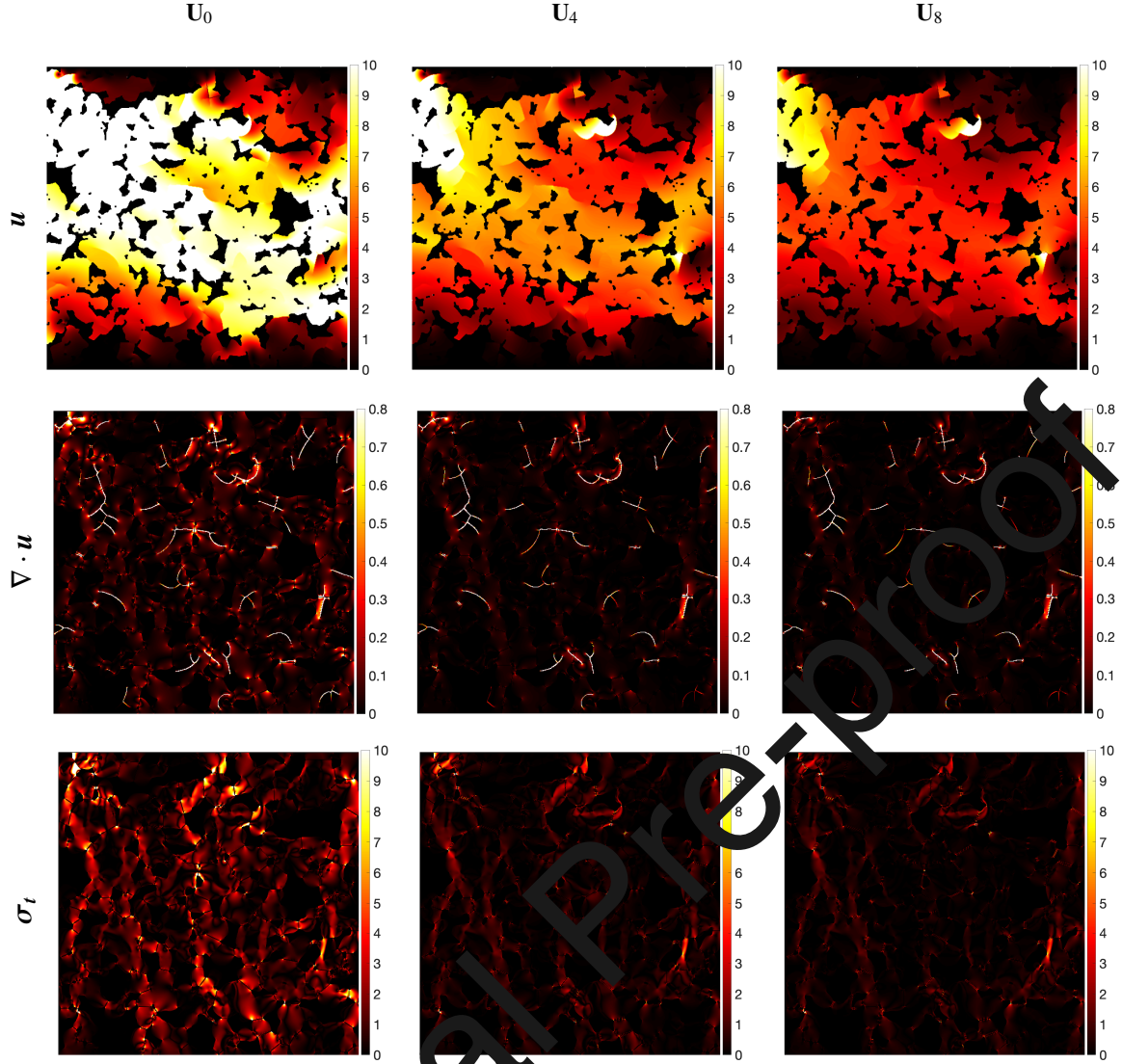


Figure 10: Pointwise errors (%) in displacement (E_p^u), volumetric strain ($E_p^{\nabla \cdot u}$), and maximum shear stress ($E_p^{\sigma_t}$) for the PLMM solutions U_0 , U_4 , and U_8 in the domain S2D-c₃. The U_0 plots correspond to Fig 7.

loading steps (or staggered iterations in phase-field models), Strategy A becomes prohibitively slow. This is because as cracks nucleate and grow within grain grids, basis functions must be recomputed, which in turn entails solving multiple local systems on the grain grids. Notice the Kronecker delta in the BCs of Eq 13. Put differently, *frequent recomputing of basis functions in an evolving-crack problem is a bad idea*. Strategy N would have been the ideal solution, because no basis updates are necessary. However, as we observed in Section 6.3 it performed poorly. This leaves Strategy C as the preferred method for solving evolving-crack problems, because basis functions are updated only when cracks cross contact interfaces, which is a far less frequent occurrence.

7.3. Computational cost

Here, we analyze the algorithmic complexity of PLMM compared to DNS, deferring a complete numerical demonstration to future work. Assume Ω_s consists of N_f fine grids (FEM nodes here), partitioned into N^g grain grids, N^c contact interfaces, and N^ζ contact grids via the decomposition algorithm in Section 4.2. For simplicity, assume also that the decomposition is *balanced*, meaning all grain grids, and separately all contact grids, consist of roughly the

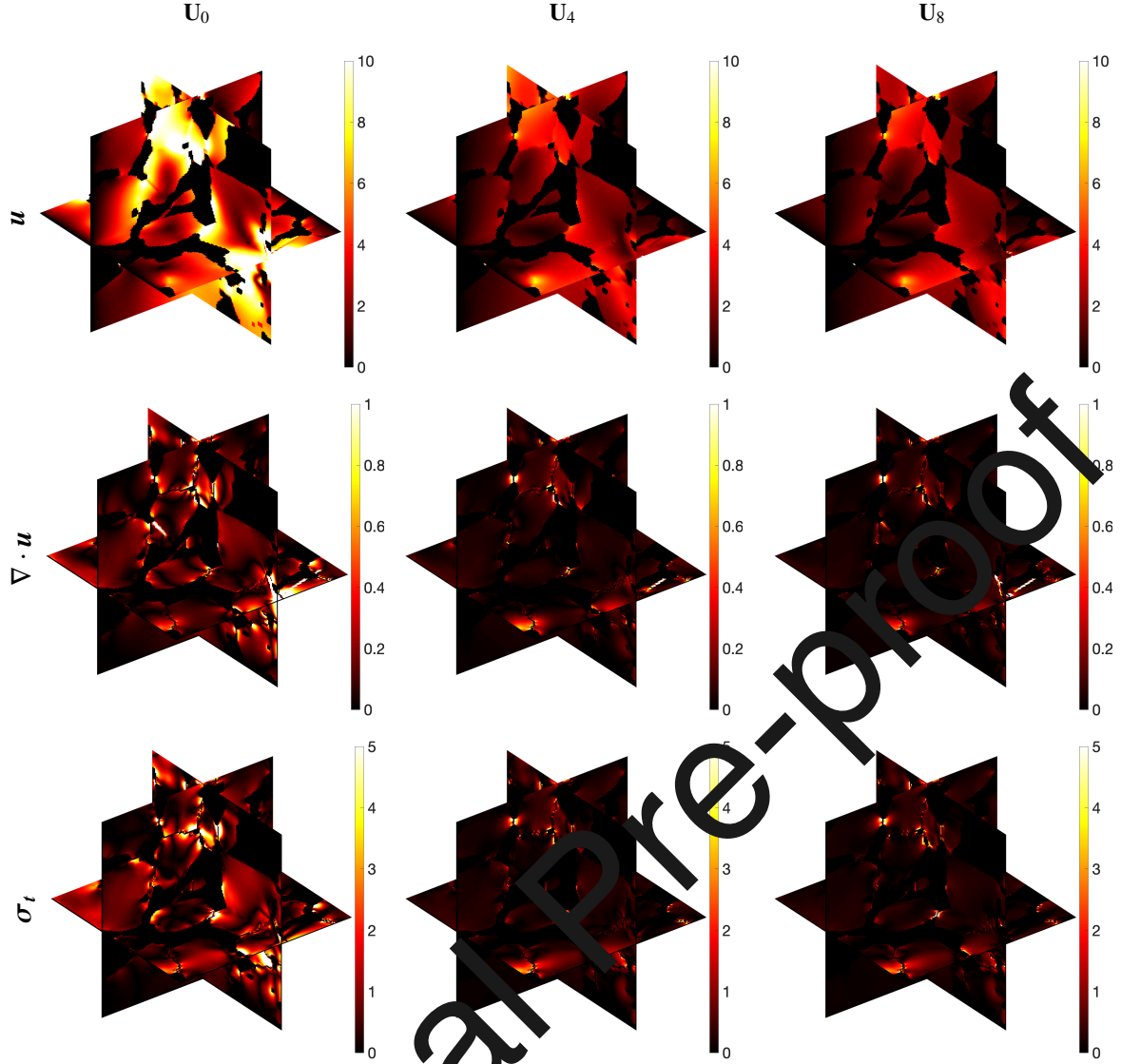


Figure 11: Pointwise errors (%) in displacement (E_p^u), volumetric strain ($E_p^{\nabla \cdot u}$), and maximum shear stress ($E_p^{\sigma_t}$) for the PLMM solutions U_0 , U_4 , and U_8 in the domain S3D-c20. The U_0 plots correspond to Fig. 8.

same number of fine grids. Let \bar{c}^{gi} denote the average number of contacts per grain grid, and f^ζ be the fraction of Ω_s covered by contact grids. As seen from Fig. 3, f^ζ is generally small (≈ 0.28 for P2D and S2D). Assume we solve all linear systems with the same solver, whose wall-clock time \mathcal{T} scales as $O(n^\beta)$, where n is the number of degrees of freedom in the system and $\beta \in (1, 3)$ is the computational complexity of the solver [71]. The wall-clock times of DNS (\mathcal{T}_S) and PLMM ($\mathcal{T}_{M_{n_b}}$), where n_b is the number of boundary iterations, can be estimated as:

$$\mathcal{T}_S = O(N^f D)^\beta \quad (33a)$$

$$\mathcal{T}_{M_0} = \mathcal{T}_{basis} + \mathcal{T}_{corr} + \mathcal{T}_{glob} + n_s (\mathcal{T}_{corr} f^F + \mathcal{T}_{glob}) \quad (33b)$$

$$\mathcal{T}_{M_{n_b}} = \mathcal{T}_{M_0} + n_b (\mathcal{T}_{cont} + \mathcal{T}_{corr} + \mathcal{T}_{glob} + n_s (\mathcal{T}_{corr} f^F + \mathcal{T}_{glob})) \quad (33c)$$

where

$$\mathcal{T}_{basis} = O(N^f D / N^g)^\beta \times (N^g \# C_{g_i}^D / P_{basis}) \quad (33d)$$

$$\mathcal{T}_{corr} = O(N^f D / N^g)^\beta \times (N^g / P_{corr}) \quad (33e)$$

$$\mathcal{T}_{cont} = O(N^f D f^\zeta / N^\zeta)^\beta \times (N^\zeta / P_{cont}) \quad (33f)$$

$$\mathcal{T}_{glob} = O(N^c D)^\beta \quad (33g)$$

The wall-clock times \mathcal{T}_{basis} , \mathcal{T}_{corr} , \mathcal{T}_{cont} and \mathcal{T}_{glob} correspond to solving the basis problem (Eq. [13], [15] or [19]), correction problem (Eq. [14], [17] or [29]), contact problem (Eq. [28]), and global problem (Eq. [26] or [30d]) in PLMM, respectively. P_{basis} , P_{corr} and P_{cont} denote the number of processors used to solve the basis, correction, and contact problems (= 1 if serial computing). Because these local problems are decoupled, the maximum number of processors that can be used in each case is $P_{basis} = N^g \# C_{g_i}^D$, $P_{corr} = N^g$, and $P_{cont} = N^\zeta$. In Eq. [33b]-c, f^F is the fraction of grain grids that contain cracks *not* captured by the basis functions; thus requiring source corrections. In DNS, the total number of unknowns is $N^f D$, whereas in PLMM, it is $(N^f + N^c)D$. The extra $N^c D$ unknowns are due to coarse-scale displacements defined at the contact interfaces, which are continuous due to Eq. [21] (i.e., factor of 2 not required). The increase in unknowns is negligible because $N^c \ll N^f$ (e.g., $N^c / N^f = 10^{-3}$ in S2D).

If Strategy A is used to compute basis/correction functions, $n_s = 0$ because no source iterations are needed, and $f^F = 0$ because cracks are included in all basis functions. Since \mathcal{T}_{glob} and \mathcal{T}_{cont} are negligible compared to \mathcal{T}_{basis} and \mathcal{T}_{corr} , we can approximate $\mathcal{T}_{M_0} \approx \mathcal{T}_{basis} + \mathcal{T}_{corr}$ and $\mathcal{T}_{M_{n_b}} \approx \mathcal{T}_{basis} + (n_b + 1) \mathcal{T}_{corr}$. The blue contour lines in Fig. [13] of

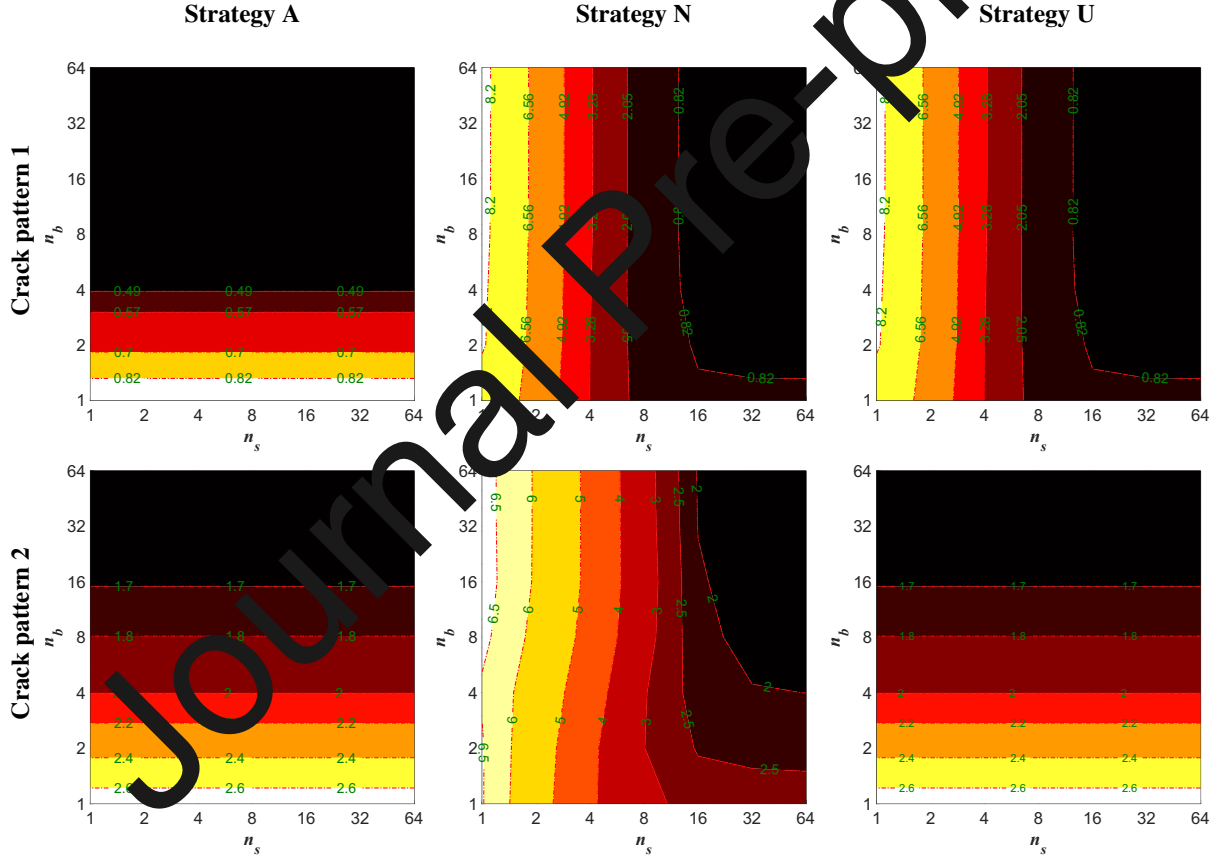


Figure 12: Relative L_2 -errors (%) in displacement (E_2^u) plotted for different combinations of source (n_s) and boundary (n_b) iterations in the P2D-c₁ and P2D-c₂ domains using Strategies A, N, and U.

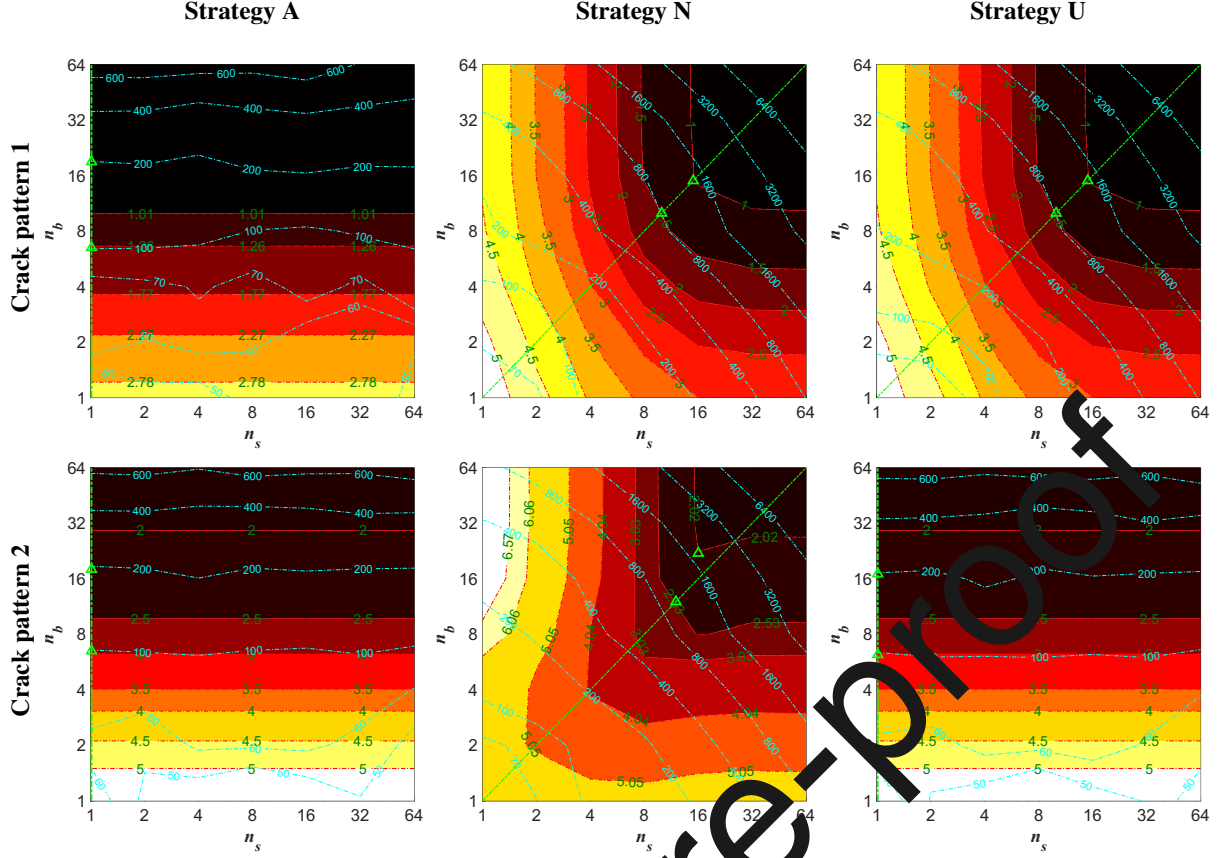


Figure 13: Relative L_2 -errors (%; red contours) in displacement (E^u) plotted for different combinations of source (n_s) and boundary (n_b) iterations in the S2D- c_1 and S2D- c_2 domains using Strategies A, N, and U. Blue contours correspond to measured wall-clock times in seconds. Optimal (n_s, n_b) -pairings are marked by green triangles. Diagonal and vertical green lines correspond to $n_s = n_b$ and $n_s = 0$, respectively; the approximate loci of optimal pairings.

measured wall-clock times for S2D- c_1 and S2D- c_2 confirm these estimates. If Strategy N or U is used, $n_s \neq 0$ and $f^F \neq 0$ and we can approximate $\mathcal{T}_{M_0} \approx (1 + n_s f^F) \mathcal{T}_{corr}$ and $\mathcal{T}_{M_{n_b}} \approx (1 + n_b)(1 + n_s f^F) \mathcal{T}_{corr}$, where we omitted \mathcal{T}_{basis} because it is a one-time cost that becomes negligible as n_s and n_b grow. The difference between Strategies N and U is that both n_s and f^F are larger for Strategy N with the possibility $n_s = \infty$ if source iterations diverge. The blue contours in Fig 13 of measured wall-clock times once again confirm these estimates.

Can n_s and n_b be chosen optimally in PLMM to satisfy a desired error tolerance at minimal computational cost? The answer is yes, and Fig 13 shows how. Focusing on Strategy N and crack pattern 1, as an example, assume our tolerance for error is $E_2^u = 1.5\%$. The corresponding contour line in Fig 13 colored red and annotated green, is the locus of all (n_s, n_b) pairings that satisfy this tolerance. If we now consider the superposed contour lines of wall-clock time, colored and annotated blue, we can determine the optimal pairing: *the point at which the blue and red contours become tangent*, marked by a green triangle in Fig 13. Similar optimal pairings can be found for different tolerances (e.g., second triangle in Fig 13 for $E_2^u = 1\%$). By connecting these optimal pairings we find that they lie roughly on the line $n_s = n_b$, which is a key takeaway. In Strategy A, optimal pairings lie on the line $n_s = 0$, shown by the vertical green lines in Fig 13. In Strategy U, optimal pairings lie on $n_s = n_b$ for crack pattern 1 (similar to Strategy N), but on $n_s = 0$ for pattern 2 (similar to Strategy A), for reasons discussed in Section 6.3.

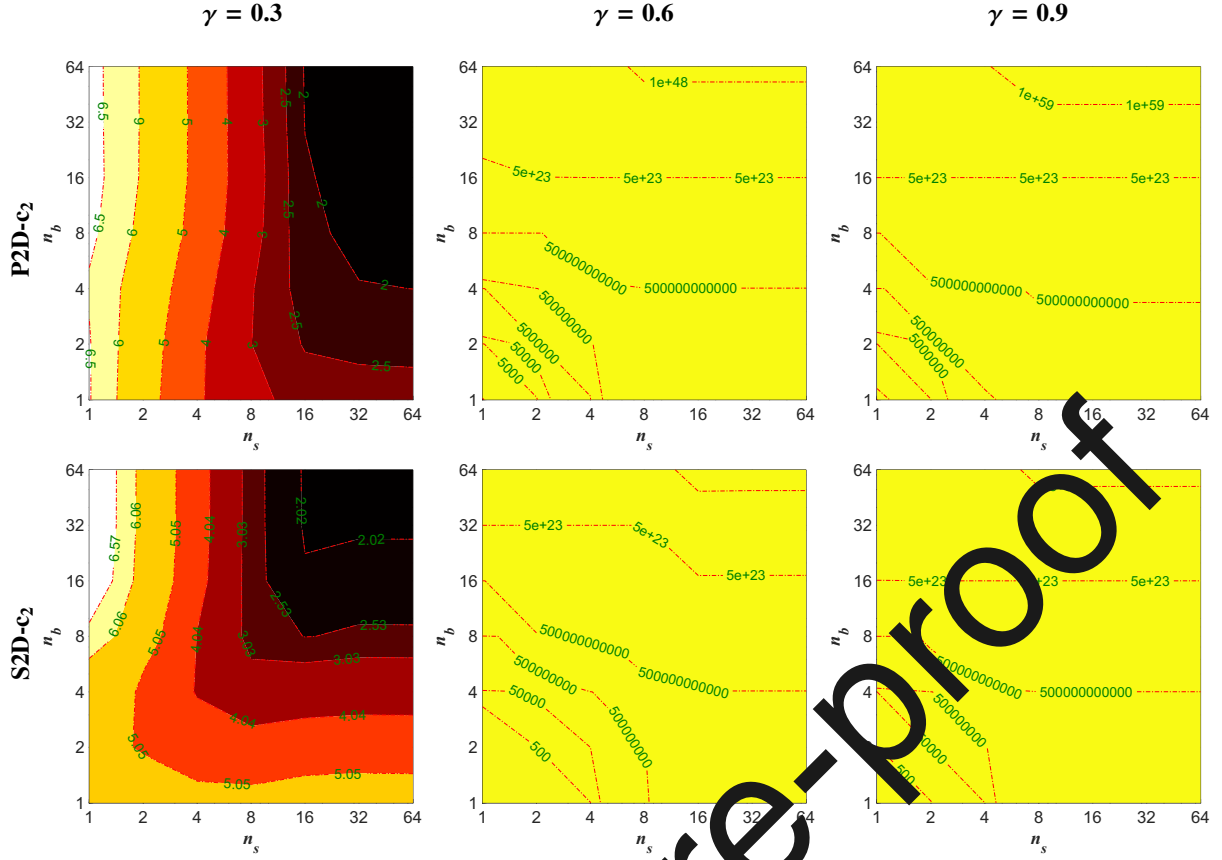


Figure 14: Relative L_2 -errors (%) in displacement (E_2^u) plotted for different combinations of source (n_s) and boundary (n_b) iterations in the P2D- c_2 and S2D- c_2 domains using Strategy N with relaxation parameters $\gamma = 0.3, 0.6$ and 0.9 .

8. Conclusion

We presented a pore-level multiscale method (PLMM) to model elastic deformation of fractured porous materials with arbitrary microstructure and crack patterns. A key feature of the proposed PLMM is that the domain decomposition need not conform to crack positions, unlike a previous formulation that had been proposed by the authors [16]. The added flexibility leads to more accurate solutions, better error control, and the absence of a need to dynamically update the decomposition when cracks evolve. We validated PLMM against single-scale direct numerical simulations (DNS) and showed it produces an accurate initial approximation (e.g., $E_s^u < 10\%$) on complex 2D and 3D geometries. We also proposed an iterative strategy to reduce the errors of this approximate solution to any desired level.

Anticipating application to evolving-crack problems in the future, three strategies for computing basis functions within PLMM were explored: (1) Strategy A, where *all* cracks are included in the basis functions; (2) Strategy N, where *no* cracks are included in the basis functions; and (3) Strategy U, where only those cracks are included that intersect at least one contact interface. If cracks are fixed, Strategy A is preferred, as it converges the fastest. If cracks evolve, Strategy U is preferred because it requires infrequent updates of the basis functions (unlike Strategy A), which is computationally more efficient. Strategy N either diverges, if not relaxed, or converges slowly. It is therefore not recommended. Future work will focus on applying PLMM to crack nucleation and growth problems. This work also opens the possibility to adapt PLMM as the linear solver for nonlinear problems involving finite strains or plasticity.

Acknowledgments

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Appendix A. Multiscale error versus fine-grid refinement

Consider a rectangular domain $\Omega = [0, 0.96] \times [0, 0.98]$ with different crack patterns, as shown in Fig. A.1. The domain is fixed at the bottom boundary and a vertical-upward displacement of $u_D = 0.001$ mm is applied to the top boundary. All material properties are the same as those in Section 6. For each domain, Eqs. 1 and 7 are solved using Strategy U of PLMM and the displacement error, E_2^u , is calculated against the single-scale DNS solution. We set the number of boundary, n_b , and source, n_s , iterations to be sufficiently large so that the PLMM solution no longer improves in accuracy. This ensures $E_b = E_s = 0$ in Eq. 32. We repeat the above process for different fine grids, refined according to $h = h_0/N_{ref}$, where h_0 is the initial fine-grid size and N_{ref} is the refinement factor. We set the phase-field length scale parameter to $l = 2h$. Fig. A.2 shows that the plot of E_2^u versus N_{ref} follows a first-order convergence behavior, regardless of the crack pattern. Together with Eq. 32, this implies $E_f \sim O(h)$.

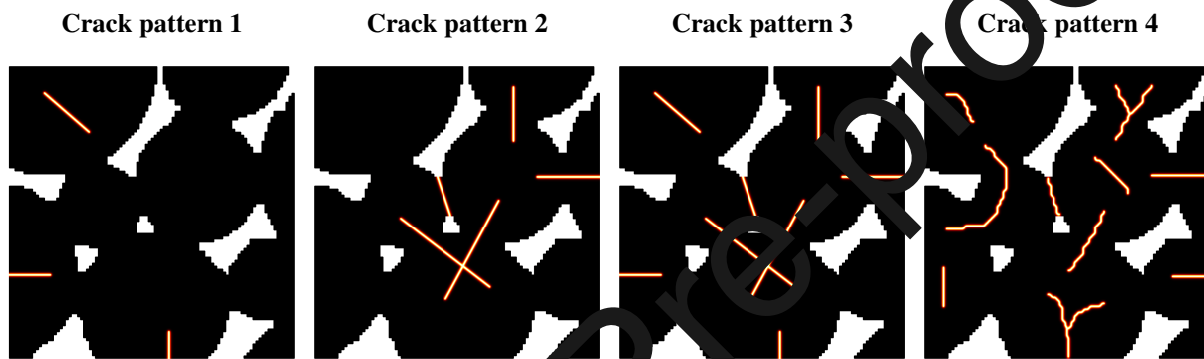


Figure A.1: Crack patterns considered in the fine-grid convergence analysis. Pattern 1: all cracks are confined to the interior of grain grids. Pattern 2: all cracks intersect at least one contact interface. Pattern 3: a combination of crack patterns 1 and 2. Pattern 4: a complex crack pattern with curved and branching cracks.

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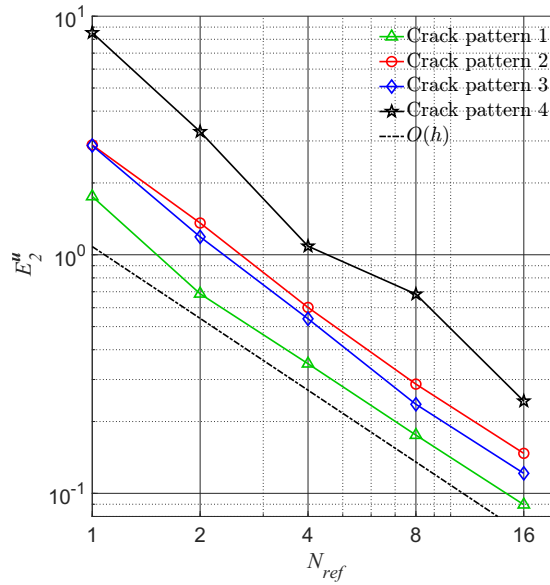


Figure A.2: Relative L_2 -errors in displacement (E_2^u) versus fine-grid refinement factors, N_{ref} for different crack patterns shown in Fig. A.1 solved with Strategy U of PLMM. At each N_{ref} , boundary, n_b and source, n_s , iterations are sufficiently large so that E_2^u no longer improves. Plotted errors represent E_f in Eq. 52 and are $O(h)$.

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Declaration of interests

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Kangan Li: Methodology, Software Development, Formal analysis, Writing - Original Draft

Yashar Mehmani: Conceptualization, Methodology, Software development, Writing - Review & Editing, Funding acquisition, Supervision

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