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An efficient matrix-free preconditioned conjugate gradient based multigrid method for phase field modeling of fracture in heterogeneous materials from 3D images

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Abstract

In this paper, we present a new and efficient strategy to perform 3D computational fracture modeling from computed tomography (CT) images. The image-based fracture modeling creates a new way to more accurately predict fracture in realistic structures. However, it is currently complex and expensive to perform such simulations. In this work, we propose to use a matrix-free type preconditioned conjugate gradient solver based multigrid method to achieve the fastest numerical performance for phase field modeling of fracture. Several specific methods are investigated to enhance the robustness and to improve the efficiency of the proposed strategy. An automatic load control strategy to control crack propagation is investigated. High performance computing is applied using a hybrid MPI/OpenMP strategy. With all of the proposed procedures, the phase field modeling of fracture can be performed in the real 3D microstructure of heterogeneous materials using tomographic images.

Keywords— Preconditioned conjugate gradient based multigrid method – Phase field fracture modeling – Heterogeneous material – Image-based simulation

1 Introduction

Industrial materials, e.g. composite materials, biological materials, architectured materials, become more and more complex both on components and structures. The complexity of these heterogeneous materials can improve their durability. But it leads to many engineering problems, e.g. failure, which is due to the limited comprehension of these materials. The most common failure mode of engineering materials and structures is fracture, which can sometimes have fearful consequences, e.g. the fracture of a bridge. Researchers pay, therefore, close attention to the prevention of crack induced failure in engineering designs. Numerical modeling of fracture can help researchers to understand and predict failure of cracked structures, e.g. cast iron, composite

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materials, human bones, concrete, for which full-scale experiments are too expensive and even impracticable.

However, modeling fracture in these complex materials is very difficult. One of the reasons is related to the real microstructure of these highly heterogeneous materials. Fortunately, experimental techniques such as X-ray tomography [1] reveal the real microstructure of these heterogeneous materials. It gives us the opportunity to analyze micro/macro interactions. Meanwhile, it confirms that there are large differences between real and theoretical (ideal) microstructures, e.g. many defects can be found in the microstructure of cast iron presented in [2] and the microstructure of fiber reinforced composite laminate illustrated in [3].

The subject of this work is to perform image-based fracture modeling using real microstructures of heterogeneous materials.

Besides the complexity of highly heterogeneous materials, the fracture modeling is itself complex for various reasons. Firstly, the crack is usually very thin compared to the structural dimension. Numerically, it can create large discontinuities with traditional methods, e.g. finite element methods. Secondly, fracture is an irreversible process, it should not heal itself. Time-dependent constraints are therefore introduced when solving such a problem. At the same time, fracture nucleation and branching can lead to more difficulties. The phase field modeling is one of the most widely used and developed damage model due to advantages such as straightforward modeling of complex crack patterns and crack branching. The variational phase-field based approach was first introduced in [4] based on [5], in which Griffith’s model ([6]) was represented by an energy minimization problem. [4] then proposed a smooth function to deal with the sharp discontinuous problem in the phase field fracture modeling. Now the phase field modeling has become very popular in the last decades. Many papers can be found both on extensions and applications of this model e.g. [7, 8, 9, 10, 11, 12, 13, 14, 15].

Many publications can be found on phase field fracture modeling, even for 3D models, e.g. [16, 7, 17, 18, 11, 13, 12]. However, few 3D works can be found for simulations using real microstructures of materials, e.g. [18]. On one hand, this is due to the complexity of real microstructure. It can create many numerical difficulties and one has to handle meshing problems due to this complexity. On the other hand, the phase field modeling is usually solved by the traditional finite element approach. One of the limitations of the standard finite element method is its high cost due to memory requirement and computational time. Several works ([19, 20, 13]) propose an adaptive mesh refinement to alleviate this bottleneck. Nevertheless, it is still difficult to perform phase field fracture modeling from real 3D images. [7] is the first paper performed image-based phase field modeling. However, they just investigated a 2D real structure, their 3D modeling is performed on a theoretical (perfect) model. Meanwhile, the cost of their simulation is very high, e.g., about 10 days for an ideal 3D model. Later, image-based phase field simulations and comparisons with in-situ experiments are presented by [17]. Highly heterogeneous image-based phase field simulations are illustrated in [18].

Based on these backgrounds and our previous work presented in [21] and [22], we propose to use a matrix free type solver to reduce memory requirement and to improve the parallel performance. This allows us to perform phase field modeling at a large scale. However, the matrix-free type solver is usually very slow to achieve convergence. We propose to apply the multigrid method to accelerate the convergence speed, which has already shown its usefulness in [21, 22] for thermal and elastic problems with billions of elements. Briefly, a matrix-free type solver based multigrid algorithm is presented in this paper to perform image-based phase field modeling of fracture in heterogeneous materials at the microscopic scale.

However, one of the limitations of multigrid methods is its ability to handle large jumps, e.g. high heterogeneity and rough solutions [23]. The adaptive nature of conjugate gradient can compensate for the shortcomings of the basic smoothing procedure in multigrid algorithms [23]. The combination of PCG methods and multigrid algorithms has been developed by researchers since the 1980’s. [24] and [25] proposed to use the multigrid method as a preconditioner of conjugate gradient methods, which is the so-called MGCG method. [26] generalized their work
10 years later. A parallel version can be found in [27]. Applications to large scale problems were solved by the MGCG in [28]. The application of MGCG methods on mixed finite element methods was investigated in [29]. On the other hand, [23] proposed to use the PCG iterator as a smoother for multigrid algorithms, which is the so-called PCGMG method. [30] combined the MGCG method and the PCGMG method to achieve a good efficiency for elasticity problems. The ability to avoid the locking effect of multigrid methods using conjugate gradient solvers was confirmed by [30]. We therefore propose to use the PCGMG method to perform the image-based phase field modeling in heterogeneous materials. This choice was made for several reasons: (1), the Jacobi iterator based multigrid method can lead to divergence when solving the mechanical part in phase field models due to the presence of cracks, which leads to huge discontinuities; (2), replacing the Jacobi smoother by a PCG iterator does not require a amount of work in terms of implementation based on our previous work; (3), a good parallel performance can therefore be ensured with minimum modifications.

The outline of this paper is the following: Section 2 briefly presents the framework of the staggered phase field modeling. Section 3 introduces the implementation of the proposed preconditioned conjugate based multigrid algorithms by using a finite element discretization. The convergence performance analysis is employed in Section 4. Several improvements are also proposed for problems with large variations in this section. The validation of the overall program and its applications are illustrated in Section 5. The snap-back behavior during the phase field modeling is also presented in this section. An automatic load-control strategy for adjusting the displacement increment is investigated for applications in this section. Conclusions are presented at the end of this paper.

2 Problem statement

Assuming domain $\Omega \subset \mathbb{R}^D$ describes a cracked solid, with its boundary $\partial \Omega$ and $D$ the space dimension. Let $\Gamma$ be the crack in $\Omega$ as presented in Figure 1a for a $D = 2$ case.

![Figure 1: Crack topology approximation in a $D = 2$ case](image)

According to [31, 6] and [5], the energy equilibrium in the cracked domain $\Omega$ can be presented as:

$$\mathcal{E}(u, \Gamma) = \mathcal{E}_u(u, \Gamma) + \mathcal{E}_s(\Gamma) = \int_{\Omega} W_u(\epsilon(u)) \, d\Omega + g_u \mathcal{X}^{D-1}$$  

where $\mathcal{E}_u(u, \Gamma)$ is the elastic energy which can be described as the integral of strain energy density $W_u(\epsilon(u))$ in the entire domain. $\epsilon = \frac{1}{2}(\nabla u + \nabla u^T)$ denotes the strain with the displacement $u$. $\mathcal{E}_s(\Gamma)$ denotes the energy required to create the crack in $\Omega$ according to the Griffith criterion in [31, 6]. It can be expressed as the product of the Hausdorff surface measure $\mathcal{X}^{D-1}$ and the
fracture toughness $g_c$ which represents the energy required to create a unit cracked surface for $D = 3$.

2.1 Phase field modeling

However, to solve a fracture problem with a sharp crack (Figure 1a) numerically, is complex. Meanwhile, crack initiation, branching or more generally topological changes of the crack are extremely difficult to model. [4] therefore proposed a smooth function instead of a sharp crack. The formulation to represent the diffuse crack illustrated in Figure 1b, is presented for example in [16]. It reads:

$$
\begin{cases}
    d - \ell_c^2 \Delta d = 0 & \text{in } \Omega \\
    d(x) = 1 & \text{on } \Gamma \\
    \nabla d(x) \cdot n = 0 & \text{on } \partial \Omega
\end{cases} \quad (2)
$$

where $d(x)$ describes the states of materials, which is defined as a crack phase field in [16], with $x \in \Omega$. $\Delta$ is the Laplacian operator. $\ell_c$ represents the actual diffuse crack thickness. Equation (2) is the Euler-Lagrange equation associated with the variational problem:

$$
d = \text{Arg} \left\{ \inf_{\Gamma_d} \gamma(d, \nabla d) \right\} \\
d \in \{d | d(x) = 1 \text{ on } \Gamma \ \forall x \in \Gamma \} \quad (3)
$$

where:

$$
\gamma(d, \nabla d) = \frac{1}{2} \ell_c^2 + \frac{g_c}{2} (\nabla d)^2 \quad (4)
$$

In the phase field modeling, $\mathcal{X}^{D-1}$ denotes the crack surface functional $\Gamma_d$. Substituting equation (1) and (3), we obtain:

$$
\mathcal{E}(u, \Gamma) = \int_{\Omega} W_u(\epsilon(u)) \ d\Omega + g_c \int_{\Omega} \gamma(d, \nabla d) \ d\Omega = \int_{\Omega} W \ d\Omega \quad (5)
$$

where the free energy $W$ reads:

$$
W = W_u(\epsilon(u)) + \frac{g_c}{2\ell_c} d^2 + \frac{g_c \ell_c}{2} (\nabla d)^2 \quad (6)
$$

To correctly compute $W_u(\epsilon(u))$, the unilateral contact formulation presented in [32] is adopted in this work. It reads:

$$
W_u(\epsilon(u)) = \begin{cases}
    \frac{1}{2} K [tr(\epsilon)]^2 + G \epsilon_{\text{dev}} : \epsilon_{\text{dev}} & \text{tr}(\epsilon) \geq 0 \\
    \frac{1}{2} K [tr(\epsilon)]^2 + g_k(d) G \epsilon_{\text{dev}} : \epsilon_{\text{dev}} & \text{tr}(\epsilon) < 0
\end{cases} \quad (7)
$$

where $K$ is the bulk modulus and $G$ denotes the shear modulus. $\epsilon_{\text{dev}} = \epsilon - \frac{1}{3} tr(\epsilon) I$ is the deviatoric part of strain tensor. $tr(\epsilon)$ is the trace of strain tensor. $\Psi^+$ and $\Psi^-$ represent the positive and negative part of energy, respectively. It reads:

$$
\Psi^+ = \begin{cases}
    \frac{1}{2} K [tr(\epsilon)]^2 + G \epsilon_{\text{dev}} : \epsilon_{\text{dev}} & \text{tr}(\epsilon) \geq 0 \\
    G \epsilon_{\text{dev}} : \epsilon_{\text{dev}} & \text{tr}(\epsilon) < 0
\end{cases} \quad (8)
$$

and

$$
\Psi^- = \begin{cases}
    0 & \text{tr}(\epsilon) \geq 0 \\
    \frac{1}{2} K [tr(\epsilon)]^2 & \text{tr}(\epsilon) < 0
\end{cases} \quad (9)
$$
The degradation function $g_k(d)$ is chosen according to [8], which reads:

$$g_k(d) = (1 - k)g(d) + k$$

(10)

where

$$g(d) = (1 - d)^2$$

The small parameter $k$ is to maintain the well-posedness of the partially broken system as proposed in [16].

The stress $\sigma$ can therefore be expressed as:

$$\sigma = K_d tr(\varepsilon) I + 2G_d \left( \varepsilon - \frac{1}{3} tr(\varepsilon) \right)$$

(11)

where if $\Psi^+ \geq \Psi^-$

$$K_d = \begin{cases} 
(1 - k)g(d) + k \cdot K & \text{if $tr(\varepsilon) \geq 0$} \\
K & \text{if $tr(\varepsilon) < 0$} 
\end{cases}$$

(12)

and

$$G_d = [(1 - k)g(d) + k] \cdot G$$

(13)

while if $\Psi^+ < \Psi^-$, $d$ is set to 0 and the initial elastic properties $K$, $G$ are recovered. The governing equations to determine the displacement field in $\Omega$ can be written as:

$$\begin{cases} 
d = 0 & \text{if $\Psi^+ < \Psi^-$} \\
\nabla \cdot \sigma = 0 & \text{in $\Omega$} \\
u = U_0 & \text{on $\partial \Omega_D$} \\
\sigma \cdot n = f_{ext} & \text{on $\partial \Omega_N$} 
\end{cases}$$

(14)

where $U_0$ is the prescribed displacement and $f_{ext}$ is the surface forces. The Dirichlet and Neumann boundary conditions are prescribed on $\partial \Omega_D$ and $\partial \Omega_N$, respectively.

According to the second law of thermodynamics, we have:

$$\sigma : \dot{\varepsilon} - \dot{W} \geq 0$$

(15)

Substituting $\sigma = \partial W/\partial \varepsilon$, we obtain:

$$-\frac{\partial W}{\partial d} \dot{d} \geq 0$$

(16)

As the crack propagation is irreversible, $\dot{d} \geq 0$. We obtain:

$$-\frac{\partial W}{\partial d} = 0$$

(17)

Considering $k \approx 0$, we can therefore obtain the governing equations to compute the phase field $d$:

$$\begin{cases} 
2(1 - d)\mathcal{H} - \frac{\dot{\Psi}^c}{\dot{\varepsilon}}(d - l^2 \Delta d) = 0 & \text{in $\Omega$} \\
d(x) = 1 & \text{on $\Gamma$} \\
\nabla d(x) \cdot n = 0 & \text{on $\partial \Omega$} 
\end{cases}$$

(18)

where $\mathcal{H}$ represents the maximum of elastic energy over history as proposed in [16]. It reads:

$$\mathcal{H}(x,t) = \max \left( \Psi^+(\varepsilon(x,t)) - \Psi^c \right)$$

(19)

with $\Psi^c = \frac{\partial W}{\partial \varepsilon}$. It is a threshold that we prescribed according to [8]. It means the damage will only appear when $\mathcal{H} \geq \Psi^c$. 

5
The overall staggered phase field modeling can be presented as follows:

<table>
<thead>
<tr>
<th>Loop on time step t</th>
</tr>
</thead>
<tbody>
<tr>
<td>• Compute displacement ( u ) with equation (14)</td>
</tr>
<tr>
<td>• Calculate ( H ) with equation (19)</td>
</tr>
<tr>
<td>• Obtain ( d ) with equation (18)</td>
</tr>
</tbody>
</table>

end loop

3 Implementation

To perform image-based phase field modeling in heterogeneous materials, we use the strategy proposed in our previous papers [21, 22] using a finite element based multigrid method. However, we will present a more efficient solver instead of the simple Jacobi iterator proposed in [21, 22].

3.1 Finite element discretization

The procedure of finite element discretization for equation (14) was proposed in [22]. In this paper, we present a detailed finite element discretization for solving equation (18). We still propose to associate one elementary node per image voxel. The chosen element type is 8-node linear hexahedron.

The weak form of equation (18) is:

\[
\int_{\Omega} \left( 2H + \frac{g_c}{\ell_c} \right) dd^* d\Omega - \int_{\Omega} g_c \ell_c \Delta dd^* d\Omega = \int_{\Omega} 2H d^* d\Omega
\]

where \( d^* \) is the test function. Applying partial integration:

\[
\int_{\Omega} \left( 2H + \frac{g_c}{\ell_c} \right) dd^* d\Omega + \int_{\Omega} g_c \ell_c \nabla d \nabla d^* d\Omega - \int_{\partial\Omega} \nabla d \cdot n d^* dS = \int_{\Omega} 2H d^* d\Omega
\]

with \( \int_{\partial\Omega} \nabla d \cdot n d^* dS = 0 \). Equation (21) is therefore:

\[
\int_{\Omega} \left( 2H + \frac{g_c}{\ell_c} \right) dd^* d\Omega + \int_{\Omega} g_c \ell_c \nabla d \nabla d^* d\Omega = \int_{\Omega} 2H d^* d\Omega
\]

which can be written in the vector form:

\[
F_d^L = F_d^R
\]

with

\[
\begin{aligned}
F_d^L &= \int_{\Omega} \left( 2H + \frac{g_c}{\ell_c} \right) dd^* d\Omega + \int_{\Omega} g_c \ell_c \nabla d \nabla d^* d\Omega \\
F_d^R &= \int_{\Omega} 2H d^* d\Omega
\end{aligned}
\]

Applying finite element discretization:

\[
d \approx \sum_{i=1}^{N} \phi_i \hat{d}_i
\]

where \( \phi_i \) denotes the shape function. \( N \) is the number of unknowns. \( i \) denotes the node id. \( \hat{d}_i \) is thus the approximation of \( d \) at node \( i \). Supposing:

\[
d^* = \phi_i
\]
Equation (24) for elementary node $j$ of one 8-node cubic element becomes:

$$
\begin{align*}
F_{Lj}^d &= \sum_e \sum_{g=1}^8 \sum_{i=1}^8 \left[ \left( 2H + \frac{g_c}{\ell_c} \right) \phi_i \phi_j \hat{d}_i + g_c \ell_c \nabla \phi_i \nabla \phi_j \hat{d}_i \right] \\
F_{Rj}^d &= \sum_e \sum_{g=1}^8 2H \phi_j
\end{align*}
$$

(25)

where $\sum_e$ is the sum over neighboring elements, $\sum_g$ is the sum over Gauss integration points, $\sum_i$ is the sum over shape function.

### 3.2 Preconditioned conjugate gradient

In the previous part, we proposed to compute the left and the right hand of equation (23) at each elementary node, instead of using typical assembled system matrices, because a matrix-free type solver is applied to reduce memory requirements as presented in [22]. However, the Jacobi solver used in [21] and [22] can not handle huge discontinuities, e.g. divergence occurs for a spherical inclusion mechanical problem with a material property contrast of 1000 (Figure 5). In the mechanical part of phase field modeling, huge discontinuities, e.g. contrast larger than $10^5$, can be found due to the presence of cracks. Instead of using the Jacobi solver, we therefore propose to use the preconditioned conjugate gradient (PCG) type relaxation in this work.

We proposed to use a diagonal preconditioner as presented in [22]. The idea is to compute the diagonal values of system matrix at each node, e.g. the preconditioner to solve equation (23) at node $i$ reads:

$$
M_i^d = \sum_e \sum_{g=1}^8 \sum_{i=1}^8 \left[ \left( 2H + \frac{g_c}{\ell_c} \right) \phi_i \phi_i + g_c \ell_c \nabla \phi_i \nabla \phi_i \right]
$$

(26)

With this technique, one does not use any matrix, which is the so-called matrix-free type finite element methods.

The generic PCG solver is presented in the following flow chart:
• Initial guess $\bar{x}$ with Dirichlet boundary conditions
• Compute preconditioner vector $M$
• $r = F_R - F_L$ with $r = 0$ on Dirichlet boundary conditions
• $ite = 1$
• while $ite \leq$ number of relaxations
  1. $z_{i,j,k}^m = r_{i,j,k}^m / M_{i,j,k}^m$
  2. update $z$ on ghost points.
  3. $rz_{new} = r^T \cdot z$
  4. $\beta = rz_{new} / rz_{odd}$ with $\beta = 0$ when $ite = 1$
  5. $p = z + \beta p$
  6. $w = F_L(p)$
  7. $pw = p^T \cdot w$
  8. $\alpha = rz_{new} / pw$
  9. $x = x + \alpha p$
 10. $r = r - \alpha w$
11. $r = 0$ on Dirichlet boundary conditions
12. $rz_{odd} = rz_{new}$
13. $ite = ite + 1$
• end

$x$ represents $u$ for mechanical equations and $d$ for phase field equations. $(i,j,k)$ indicates the node ID. $m = 1, 2, 3$ for the mechanical part and $m = 1$ for the phase field equation. $r$ represents the residual which is the difference of the right and the left hand side. $p$, $w$ and $z$ are temporary vectors. $\alpha$ and $\beta$ are temporary coefficients required for the PCG method. The ghost points mentioned at the second step of the loop, are created to ensure the presence of neighbors in order to perform the integration over the entire support of the nodes of a given MPI domain. They are indispensable to apply the matrix free type FEM, after the MPI domain decomposition.

3.3 Multigrid methods

Besides the PCG solvers, intergrid operators are required in multigrid algorithms. Different from the intergrid operators presented in [21] and [22], specific treatment is required to perform phase field modeling.

Equation (18) indicates that, $H$, $\ell_c$ and $g_c$ on the coarse grid are required to perform multigrid methods. In this work, $\ell_c$ is considered as a constant, which has the same value on all grids. For $g_c$, which is assigned to elementary nodes, homogenization techniques are used to obtain coarse grid values as presented in [21]. Besides $\ell_c$ and $g_c$ on coarse grids, $H$ is also required on coarse grids. As illustrated in equation (19), $H$ is computed at each Gauss integration point. Furthermore, $H$ is a history variable that is to be stored over time steps. However, storing $H$ at each Gauss integration point is too expensive. e.g., for $N^3$ 8-node cubic elements, it is $8$ times more expensive to store $H$ at the Gauss integration point than storing a single representative value for each element. To save memory space, we thus propose to store $H$ at each element. It
is obtained by a simple average of 8 values at Gauss integration points in each element. The integration of the initial phase field problem is thus approximated but, it will be demonstrated in the examples presented in the last Section that it does not introduce any bias in the overall problem resolution. For the $\mathcal{H}$ on coarse grids, we proposed to generate them by a simple average of the corresponding elements from the finest grid. It can be computed recursively as presented in [21].

With $\mathcal{H}$, $\ell$ and $g_c$ on coarse grids, the PCG relaxations can be executed without any obstacles for equation (18). To transform the correction and the residual between grids, standard trilinear prolongation and restriction operators are used.

To solve equation (14), equations (12) and (13) have to be updated at each time step, due to the change of $d$.

As mentioned above, one uses the strategy of one voxel per elementary node. Initial material properties are therefore assigned to each node. However, in finite element methods, what we use is the material property at the Gauss integration points. One proposes to perform an interpolation with the above mentioned shape function. For example for the shear modulus, it reads:

$$G^g = \sum_{i=1}^{8} \phi_i G_i$$

(27)

where $G^g$ denotes the shear modulus at a Gauss integration point and $G_i$ is the nodal shear modulus. $i$ is the node id. However, when we apply it to equation (13), it becomes a little more complicated due to the presence of $d$. The $d$ at Gauss integration points must firstly be computed:

$$\begin{cases}
    d^g = \sum_{i=1}^{8} \phi_i d_i & \Phi^+ \geq \Phi^-
    \\
    d^g = 0 & \Phi^+ < \Phi^-
\end{cases}$$

(28)

where $d^g$ denotes $d$ at Gauss integration points. $d_i$ is the nodal value. $G$ is then computed at Gauss integration points by equation (27). Finally, $G_d$ at Gauss integration points, reads:

$$G_d^g = [(1-k)(1-d^g)^2 + k] \cdot G^g$$

(29)

Applying the above strategy on the finest grid is straightforward and mandatory. However, to obtain $K_d$ and $G_d$ on coarse grids requires several specific treatments.

As demonstrated in equations (12) and (28), one must first compute $tr(\epsilon)$, $\Phi^+$ and $\Phi^-$ to define $K_d$ and $G_d$ at each Gauss integration point. To compute $tr(\epsilon)$, $\Phi^+$ and $\Phi^-$, the displacement $u$ is indispensable. However, in phase field modeling, the structure is globally under traction or shearing, i.e. $tr(\epsilon) \geq 0$. Nevertheless, compression, i.e. $tr(\epsilon) < 0$, can be presented in several small zones, or even several nodes on the finest grid. But, on coarse grids, the displacement has been homogenized, one may not find compression on the coarse grids. Using the displacement on coarse grids to obtain $tr(\epsilon)$, $\Phi^+$ and $\Phi^-$, can therefore not guarantee a reasonable representation of the original problem on coarse grids. It can cause slow convergence and even divergence in multigrid algorithms. By consequence, we propose to compute these criteria on Gauss integration points at the finest grids. For the $K_d$ and $G_d$ on coarse grids, the following strategy is proposed:

- Calculate the value of $K$, $G$ and $d$ at Gauss integration points on the finest grid.
- Compute $tr(\epsilon)$, $\Phi^+$ and $\Phi^-$ at these Gauss integration points on the finest grid.
- Obtain $K_d$ and $G_d$ at these Gauss integration points on the finest grid.
- Calculate the average of $K_d$ and $G_d$ of all Gauss points in each element. These averages, i.e. $K_e$ and $G_e$, represent elementary material property on the finest grid.
• Compute $K_e$ and $G_e$ in each element on coarse grids with homogenization techniques presented in [22].

The idea of the above strategy is to use the averaged $K_d$ and $G_d$ to transfer Gauss integration point material property to element. The coarse grid material property is then computed directly from $K_e$ and $G_e$ on the finest grid. The best representation of the original problem is achieved. Note that our objective is to solve the problem on the finest grid, the existence of coarse grids is to improve convergence rate. That is why on coarse grid, we do not seek an exact solution of the original problem but making several modifications to achieve the best convergence.

The prolongation and restriction operators used for solving equation (14) are tri-linear standard operators.

For details of the multigrid algorithm, please refer to our previous work [21, 22]. Here, we just take some notations. The full multigrid (FMG) cycle and V-Cycle are used in this work. Figure 2 presents a 3-level FMG scheme. One performs $\nu_0$ relaxations on level 1, $\nu_1$ relaxations on each level going up and $\nu_2$ relaxations on each level going down. $n_{cy}$ V-Cycles are used on each level. A trilinear interpolation of the solution of level $l$ is applied to obtain the initial solution of level $l + 1$. The V-Cycle program for the PCGMG method is illustrated in the following flow chart:

1. Compute the coarse grid material property.
2. Carry out $\nu_2$ relaxations with the PCG solver on the finest level $l$.
3. Inject the solution and restrict the residual to level $l – 1$.
4. Perform $\nu_2$ relaxations on level $l – 1$.
5. Repeat steps 3,4 until the coarsest grid $l = 1$.
6. Perform $\nu_0$ relaxations on level 1.
7. Prolong the correction to the next finer level and perform $\nu_1$ relaxations on this level.
8. Repeat step 7 until the finest level.
9. Loop step 2 – 7 until obtaining the required relative error.
10. Output results

Figure 2: FMG cycles
3.4 Overall program

The above sections introduce each part of the phase field modeling. In this section, the entire program to perform the phase field modeling in heterogeneous materials is presented. Besides of the numerical efficiency, the hybrid MPI/OpenMP is also applied to achieve the best performance both on memory and computational time. The entire program is described in the following flow chart.

- Create MPI topology
- Input images and obtain \( g_c, K \) and \( G \) on the finest grid.
- Calculate \( g_c \) on all coarse grids
- \( t = 0 \) and initialization of vectors
- while \( t \leq \) number of time step
  1. if \( (t=0) \)
     (a) Compute material properties, i.e. \( K_e \) and \( G_e \), depend on \( d \) on all grids
     (b) Compute \( U_e \) where \( \max(W(x)) = \Phi_c \)
     (c) \( U_t = U_e \)
  2. else
     (a) Calculate prescribed displacement \( U_t = U_t + \delta U \)
     (b) Compute \( u_t \) always with a prescribed displacement \( U_{unit} \) but a new \( d \)
     (c) Compute \( H \) with \( \frac{U_t}{U_{unit}} u_t \)
     (d) Compute \( d \) with the new \( H \)
     (e) output \( d \)
  3. end if

From the above algorithm, we can find that, in step 2.(c), instead of computing the displacement with the updated \( d \) and \( U_e \), we propose to always compute the solution \( u_t \) with the same boundary condition \( U_{unit} \) and an updated \( d \). Because equation (14) is a static linear elastic problem at each time step, its solution \( u_t \) is proportional to the prescribed displacement. With this strategy, one can minimize changes in the displacement solution on going to the next time step. It can lead to a faster convergence speed when using an iterative solver.

4 Performance analysis

In this section, the convergence performance of the proposed method is analyzed. Some improvements are also investigated. The parallel performance is also demonstrated at the end of this section.

4.1 Convergence performance of the proposed algorithm

The performance of the PCGGMG method is firstly analyzed by solving a spherical inclusion problem in a linear elastic domain \( \Omega \) as illustrated in Figure 3. The cube size is \( L \). \( \Omega \) is
discretized by $128^3$ elements. A 6-level multigrid algorithm is applied. The element size on level $l$ is 2 times larger along each direction than on level $l+1$. For this first example, we propose to use simple trilinear restriction and prolongation operators to ensure the transfers between grids. The material properties on coarse grids are computed by a simple average of the material property at corresponding nodes (or element, depending on parameters) from the finest grid. The matrix and the inclusion have the same Poisson ratio $\nu = 0.29$. The Young’s modulus of the matrix is $E_M = 233.43$ GPa. Supposing the Young’s modulus of inclusion $E_I$ is smaller than $E_M$, the contrast is defined by:

$$c = \frac{E_M}{E_I}$$

The following boundary conditions are prescribed:

$$\begin{cases}
    u_z = U_o & \text{on } z = L. \\
    u_z = 0 & \text{on } z = 0. \\
    u_x = 0 & \text{at node } \left(\frac{L}{2}, 0, 0\right) \\
    u_y = 0 & \text{at node } \left(0, \frac{L}{2}, 0\right)
\end{cases}$$

To analyze the convergence performance, the relative residual $Rr$ is computed by the following equation:

$$Rr = \frac{r^T \cdot r}{F_R^T \cdot F_R} \times 100\% \quad (30)$$

Figure 4 illustrates the convergence performance for the single level PCG solver and the PCGMG method with a material contrast $c = 1000$. The relative residual to be reached is set to be: $Rr \leq 10^{-6}$. A 2-1 V-Cycle is used for the PCGMG solver, which means $\nu_2 = 2$ and $\nu_1 = 1$. Supposing the cost of the transform between grids is negligible, the total cost of a 5 V-Cycle PCGMG is equivalent to 17 relaxations on the finest grid. On the other hand, to achieve the required $Rr$, the single level solver takes 445 relaxations, which is 26 times more expensive than the PCGMG solver.
Figure 4: Convergence of single level PCG and PCGMG algorithms

Figure 5 presents the convergence performance of the Jacobi solver based multigrid and PCG based multigrid for different material property contrasts. From this Figure, we can conclude: (1), A large contrast, e.g. $c = 1000$ can lead to non-convergence with the Jacobi based multigrid method. (2), PCGMG is more efficient (faster convergence) than Jacobi based multigrid even for small contrast cases, e.g. $c = 10$. (3), PCGMG is robust even when the contrast is closed to the numerical precision, e.g. $c = 10^{12}$.

To analyze the locking effect in multigrid methods, the Poisson ratio $\nu$ is set to be 0.49 with $c = 1000$. Figure 6 illustrates the evolution of the relative residual $Rr$ of PCGMG algorithms. The convergence speed is less than the problem with $\nu = 0.29$, but it converges. With the Jacobi based multigrid algorithms, we obtain a divergence from the first relaxation onward.

According to the performance analysis carried out on the spherical inclusion problem, we confirmed that: (1), PCGMG can handle problems with large material property variations which is the case in phase field models; (2), PCGMG is able to overcome the locking effects; (3), PCGMG is much more efficient than the Jacobi solver based multigrid algorithm; (4), The strategy to use a PCGMG algorithm is much more efficient than a single level PCG solver.

However, as presented in [21] and [22], the spherical inclusion is a simple problem with a regular and symmetric geometry. The real microstructure of heterogeneous materials is much more complicated than a spherical inclusion. A standard multigrid method may not be so efficient for handling complex microstructures. A homogenization technique was therefore proposed in [21] and [22] to compute material properties on coarse grids. It can guarantee a reasonable representation of the finest grid problem on coarse grids. Here, we use the real tomographic image of nodular graphite cast iron to analyze the convergence performance of such a proposed homogenization technique for solving a problem with a stress singularity.

As presented in Figure 7, a region of interest with $257^3$ voxels is taken from the image of graphite cast iron obtained by [2]. The voxel size is $5.06 \, \mu m$. The strategy to obtain the material property from the gray level in the image is presented in [22]. The material property of each material illustrated in Table 1 is applied. The convergence performance is analyzed with an initial crack.

In this work, we propose to work in voxel. The size of the entire domain $\Omega$ is $256^3$ voxels. $256^3$ 8-node cubic elements are used to discretize $\Omega$. The initial crack is prescribed as:

$$d = 1 \text{ when } x < 25.6 \& z = 127, 128, 129$$
Figure 5: Convergence performance of Jacobi based multigrid and PCG based multigrid for different material property contrasts.

Table 1: Material properties in cast iron

<table>
<thead>
<tr>
<th>Material</th>
<th>E/GPa</th>
<th>$\nu$</th>
<th>$g_c$/kN·mm$^{-1}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Iron</td>
<td>210</td>
<td>0.2</td>
<td>$1.73 \times 10^{-3}$</td>
</tr>
<tr>
<td>Graphite nodules</td>
<td>21</td>
<td>0.3</td>
<td>$1.8 \times 10^{-4}$</td>
</tr>
</tbody>
</table>

which is illustrated in Figure 8. The following boundary conditions are applied:

\[
\begin{align*}
    u_z &= 1 \quad \text{on } z = 256 \\
    u_z &= 0 \quad \text{on } z = 0 \\
    u_x &= 0 \quad \text{at node } (128, 0, 0) \\
    u_y &= 0 \quad \text{at node } (0, 128, 0)
\end{align*}
\]

The small parameter $k$ is set to $10^{-5}$, $\ell_c = 2h$ with element size $h$. A 7-level multigrid algorithm is applied. The element size on level $l$ is 2 times larger than on level $l + 1$. V-Cycles with $\nu_2 = 2$ and $\nu_1 = 1$ are used to solve equation (18). However, $\nu_2 = 4$ and $\nu_1 = 2$ are applied for
solving equation (14) due to higher complexity. A trilinear restriction and prolongation operator is used for solving both equations. A problem is supposed to be converged when $Rr \leq 10^{-6}$. The problem is solved by using different strategies for estimating material properties on coarse grids. A first simulation is performed in which material properties on coarse grids are obtained by averaging material properties at corresponding nodes from the finest grid. An other strategy is to use the advanced technique proposed in [21] and [22] to recursively compute material properties on coarse grids.

Figure 9a shows that the advanced technique may not give a significant improvement when solving equation (18). Because the best performance is already achieved with a simple average. But for equation (14), as presented in Figure 9b, the advanced technique can distinctly improve the convergence performance, i.e. the simulation is 2 times cheaper. That is because a simple average of material properties on coarse grid can not guarantee a reasonable representation of the original problem for such a complex case. The result is confirmed with the proposal in [22].

4.2 Parallel performance analysis

A hybrid MPI/OpenMP code is built based on our previous work [21] and [22]. The detailed performance analysis is already presented in [21] for thermal problems. To be rigorous, a brief
Figure 7: The gray level (GL) in image of cast iron

Figure 8: The prescribed initial crack. The initial crack is red and graphite nodules are green

parallel performance analysis is proposed herein. Since the analysis of the entire phase field simulation until failure is too expensive, we propose to analyze it on the previous nodular
Figure 9: Convergence performance of PCGMG algorithms with different material properties on coarse grids for both phase field problem and balance of momentum problem.

graphite cast iron problem with a stress singularity. This problem with a stress singularity can be considered as a time step of phase field modeling: one resolution of the phase field problem plus one resolution of the mechanical problem. The computations are performed on the supercomputer Liger of ICI at Ecole Centrale de Nantes. Each processor of Liger computer consists in 12 cores where OpenMP and MPI can be used. Each node has two sockets or processors. Between nodes, MPI is inevitable. Between sockets, MPI is suggested since OpenMP suffers from poor data access patterns when using two sockets. As a consequence of this computer architecture, the number of OpenMP per MPI task is limited to 12. On Liger computer, each node has 128 GB memory space. With the help of the matrix free strategy, 128 GB is enough for solving on one single core the largest numerical example, i.e. $256^3$ elements, presented in this work.

To analyze the parallel performance, the wall time is obtained with different configurations (see in Table 2) for the same simulation. Figure 10 shows the speedup as a function of the number (NB) of cores used on this $256^3$ elements problem. Within 100 cores, the speedup is almost optimal. When using 384 cores, an efficiency of about 90% can be found, which means the wall time is reduced to 150 s instead of 51567 s with one core. This efficiency confirms to the result obtained in [21].

Table 2: Hybrid setting

<table>
<thead>
<tr>
<th>NB of MPI</th>
<th>NB of OpenMP</th>
<th>Cores used</th>
<th>Nodes used</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>8</td>
<td>3</td>
<td>24</td>
<td>1</td>
</tr>
<tr>
<td>16</td>
<td>3</td>
<td>48</td>
<td>2</td>
</tr>
<tr>
<td>64</td>
<td>1</td>
<td>64</td>
<td>4</td>
</tr>
<tr>
<td>32</td>
<td>3</td>
<td>96</td>
<td>4</td>
</tr>
<tr>
<td>32</td>
<td>6</td>
<td>192</td>
<td>8</td>
</tr>
<tr>
<td>64</td>
<td>6</td>
<td>384</td>
<td>16</td>
</tr>
</tbody>
</table>

In this work, we do not show the efficiency with 1000 cores. This is because the largest numerical example presented in this work has only $256^3$ elements. 384 cores are enough to solve such problem. Using more cores can slow down the speedup. Meanwhile, for an entire phase field modeling with such structure, if we have 1000 time step, it will take less than 42 hours. Note that the first time step is the most expensive one. However, if we do not use parallel
computing, the computational time will be 596 days (almost two years).

5 Validation and applications

In this section, the validation of the proposed strategy is presented. Two applications are then presented to show the ability of the proposed strategy.

5.1 Validation with three dimensional single notched plate

To validate the proposed strategy, a typical crack propagation in a squared shape part with homogeneous material under a tensile loading is illustrated. The three-dimensional mode-I tension test has already been presented in [16] and [10]. However, only the crack pattern can be found in these two papers, the 3D load–displacement curve is not illustrated, which does not allow us to perform a quantitative comparison. In this work, we therefore take the 2D example parameters from the literature ([16, 10, 15]) and try to find an agreement between the 2D and 3D examples.

The dimension of the part and the initial crack are presented in Figure 11. The Young’s modulus is set to be 210 GPa. Its Poisson ratio equals to 0.3. The fracture toughness $g_c$ is set to $2.7 \times 10^{-3} \text{kN}\cdot\text{mm}^{-1}$. The domain is discretized by $256 \times 64 \times 256$ elements. The element size is about $h = 0.0039 \text{mm}$. $\ell_c$ is supposed to be $0.0078 \text{mm}$, which is twice the element size. The small parameter $k$ equals to $10^{-5}$. 7 levels are used for the multigrid algorithm. The element size of level $l + 1$ is two times smaller than on level $l$. 192 cores are used to perform this simulation.

![Hybrid Performance](image-url)

Figure 10: The parallel performance of the hybrid MPI/OpenMP strategy
The following boundary conditions are applied:

$$\begin{align*}
    \begin{cases}
        u = 0 & \text{on } z = 0 \text{ mm} \\
        u = \{0, 0, Ut\} & \text{on } z = 1 \text{ mm}
    \end{cases}
\end{align*}$$

As mentioned above, the displacement is always computed with the same prescribed displacement $U_{\text{unit}}$ but a new $d$. This $U_{\text{unit}}$ equals to 1 mm in this example. The following increment is applied for this case:

$$\begin{align*}
    \delta U = 1e^{-5} & \quad Ut \leq 5e^{-3} \text{ mm} \\
    \delta U = 1e^{-6} & \quad Ut > 5e^{-3} \text{ mm}
\end{align*}$$

At each time step, the solution is supposed to be converged when the relative residual $R_r \leq 10^{-6}$. The same material parameters are used according to the 2D single edge notched tensile test presented in [16, 10, 15]. From the above time step displacement increment, we can find that in this example, the threshold for $\mathcal{H}$ is not applied, i.e. $\Phi_c = 0$, for comparing with the results from literature.

Figure 11: Geometry and initial crack

Figure 12 illustrates the crack pattern at different loading step. It confirms to the 3D results presented in [16] and [10]. From this Figure, we find that the computed crack thickness is smaller than the thickness prescribed to the initial crack. This is because $d = 1$ are prescribed on three layers of nodes to introduce the initial crack in the model. But $l$ is set to $2h$, the computed crack thickness is therefore smaller.

To perform a quantitative 2D–3D comparison, we propose to compute the reaction force $F$ per unit width along the $Z$ direction. It means to divide the reaction force by the width of the domain, i.e. 0.25 mm. The load–displacement curve should then be similar to 2D regardless of 3D effects. Figure 13 illustrates the reaction force $F$ on a unit width along $Z$ direction as a function of the prescribed displacement $Ut$. A maximum reaction force $F_{\text{max}}$ of 0.66 kN/mm can be found in this case. The result is similar to the 2D load–displacement curve presented in [16, 10] and [15].
Figure 12: Evolution of crack pattern at different loading step: The isosurface of phase field $d$ with a value of 0.99. The left crack pattern is obtained at $Ut=0.005591$ mm, the middle one is obtained at $Ut=0.006$ mm, the right one is obtained at $Ut=0.006273$ mm.

5.2 Crack initiation and propagation in the structure with a spherical inclusion

In this section, the proposed strategy is applied for the modeling of crack initiation and propagation in a structure with a spherical inclusion as presented in Figure 3. The size of the cube-shaped structure is $128 \times 128 \times 128$ mm$^3$. Properties of each material are given in Table 3. One supposes that the crack resides only in the matrix. The $g_c$ in the inclusion is therefore supposed to be 100 times greater than in the matrix. $126^3$ 8-node cubic elements are used to discretize the cube with the element size $h = 1$ mm. The Multigrid method uses 6 levels. At time step $t = 0$, FMG cycles are performed. However, V–cycles with $\nu_2 = 2$ and $\nu_1 = 1$ are applied for the other time steps. It is because from one time step to the next time step, the solution should not change a lot. A good initial solution is already obtained. V–Cycles are therefore the best choice. $\ell_c = 2h$ is applied. 192 cores are used to perform this simulation. The following boundary conditions are prescribed:

$$
\begin{cases}
  u_z = U_t & \text{on } z = 128 \text{ mm} \\
  u_z = 0 & \text{on } z = 0. \\
  u_x = 0 & \text{at node (64,0,0) mm} \\
  u_y = 0 & \text{at node (0,64,0) mm}
\end{cases}
$$

At each time step, the solution is supposed to be converged when the relative residual $Rr \leq 10^{-6}$. At $t = 0$, $U_e = 0.138$ mm is obtained.

For the displacement increment $\delta U$, in this work, we propose an automatic load control strategy to better manage the crack propagation. As is well-known, the basic hypothesis of the staggered phase field is that: the solution increment shall be sufficiently small and the system shall be stable to guarantee a good approximation of the monolithic problem. It means that $d$ should not change a lot from one time step to the next one. We therefore propose to compute
the maximum value of the variation of $d$ from time step $t$ to the time step $t + 1$ over the entire domain. It reads:

$$\delta d_{\text{max}} = \max(d_{t+1}^i - d_t^i)$$

where $d_t^i$ and $d_{t+1}^i$ are the nodal values of $d$ over the entire domain at time step $t$ and $t + 1$, respectively. $\delta d_{\text{max}}$ thus estimates the maximum value of the variation of $d$ over the entire domain. The automatic time step control system is constructed based on $\delta d_{\text{max}}$. The principle is to increase $\delta U$ when $\delta d_{\text{max}}$ is too small, and to reduce $\delta U$ when $\delta d_{\text{max}}$ is too large.
For this spherical inclusion problem, the following guidelines are proposed as a first attempt:

\[
\delta U^{t+1} = \begin{cases} 
2\delta U^t & \delta d_{\text{max}} \leq 0.01 \\
\delta U^t & 0.01 < \delta d_{\text{max}} < 0.1 \\
0.5\delta U^t & \delta d_{\text{max}} \geq 0.1 \\
0.01U_e & \delta U^t \geq 0.01U_e
\end{cases}
\]

where \( \delta U^t \) denotes the value of \( \delta U \) at time step \( t \), with \( \delta U^1 = 0.01U_e \) at \( t=1 \). To avoid a too large displacement increment, the maximum value of \( \delta U \) is limited at \( 0.01U_e \).

A fully cracked structure as presented in Figure 14b is obtained at time step \( t = 91 \). The top fracture pattern confirms the result presented in [16]. Only one crack pattern can be found in [16], that is because they used irregular meshes and a refinement only on the top. It is well-known that strain concentration can be found on the two poles of the sphere. Multicrack is expected with this symmetric structure without local defects. The Z direction external force versus the displacement response is illustrated by the red solid line in Figure 15.

![Figure 14: Displacement increment applied and the fully broken structure](image)

The above result seems to be correct. However, analyzing Figure 14a, one finds that the crack continues to propagate even when \( \delta U \) becomes very close to 0, i.e. \( 10^{-15} \) mm. Meanwhile, with only 91 time steps, the crack initiation and propagation are obtained even with a such small displacement increment. It means that the crack propagation might be unstable and the crack propagates too fast.

As a consequence, the stability of the above system at each time step is suspect. The snapback behavior during the crack propagation proposed in [33] and [34] is therefore considered. The idea is to allow a decrease for the prescribed displacement during crack propagation, i.e. applying negative \( \delta U \). The following automatic control strategy is then proposed. For \( \delta U^t > 0 \)

\[
\delta U^{t+1} = \begin{cases} 
2\delta U^t & \delta d_{\text{max}} \leq 0.01 \\
\delta U^t & 0.01 < \delta d_{\text{max}} \leq 0.05 \\
0.2\delta U^t & 0.05 < \delta d_{\text{max}} \leq 0.15 \\
0.1\delta U^t & \delta d_{\text{max}} > 0.15 \\
0.01U_e & \delta U^t \geq 0.01U_e \\
-10^{-9} & \delta U^t \leq 10^{-9}
\end{cases}
\]
For $\delta U^t < 0$

$$\delta U^{t+1} = \begin{cases} 
0.5\delta U^t & \delta d_{\text{max}} \leq 0.01 \\
\delta U^t & 0.01 < \delta d_{\text{max}} \leq 0.05 \\
5\delta U^t & 0.05 < \delta d_{\text{max}} \leq 0.15 \\
10\delta U^t & \delta d_{\text{max}} > 0.15 \\
-0.01U_e & \delta U^t \leq 0.01U_e \\
10^{-9} & \delta U^t \geq -10^{-9}
\end{cases}$$

with $\delta U^1 = 0.01U_e$ at $t=1$.

Figure 17 illustrates the crack at different time steps with automatic load control. Here, 421 time steps are required to reach complete failure. The failure starts at the two poles of the sphere. It is consistent with zones where there are stain concentration on linear elastic problem as obtained in [22]. A significant difference can be found at the final failure between strategies with and without snap-back, especially for the top crack. Note that the non-symmetric crack patterns for the top and the bottom crack are consistent with the symmetry break introduced by the boundary conditions detailed above. The blue dashed line in Figure 15 shows a typical snap-back behavior. It is obtained that the maximum value of the external force is almost the same. It also confirms that during the macroscopically elastic part, the structure is stable. However, once exceeding this phase, the structure becomes unstable and the snap-back behavior is therefore obtained. Figure 16 presents the evolution of $\delta d_{\text{max}}$ as a function of the time step $t$. To simplify the comparison, the two curves are plotted in the same X scale. It is obvious that without snap-back the $\delta d_{\text{max}}$ remains large after crack initiation, even the load increment vanishes. However, with snap-back, only one pic in $\delta d_{\text{max}}$ can be found at the end of crack propagation. To be quantitative, the average of $\delta d_{\text{max}}$ over time steps is calculated for these two strategies. 0.20 is obtained for the strategy without snap-back. For the strategy with snap-back, one obtains 0.06. It confirms that the automatic load control strategy allowing snap-back can guarantee a minimum variation of $d$ for each time step. It thus legitimates the staggered resolution strategy of the phase field problem.

5.3 Crack initiation and propagation in cast iron

In this section, the crack propagation is studied in a real image of cast iron as presented in Figure 7. The geometry and material properties were detailed in Table 1. The finite element discretization strategy and the multigrid parameter setting are the same for the cast iron example presented in Section 4. 384 cores are used to perform this simulation. Here, we work with a voxel size of 5.06 $\mu$m. Different from Section 4, here, the phase field modeling is performed without an initial crack. It allows one to analyze the crack initiation and propagation. The following boundary conditions are prescribed:

$$\begin{align*}
    u_z &= Ut & \text{on } z &= 256 \\
    u_z &= 0 & \text{on } z &= 0 \\
    u_x &= 0 & \text{at node } (128, 0, 0) \\
    u_y &= 0 & \text{at node } (0, 128, 0)
\end{align*}$$
The above mentioned automatic load control strategy is applied. It reads: For $\delta U^t > 0$

$$
\delta U^{t+1} = \begin{cases} 
10\delta U^t & \delta d_{\text{max}} < 0.0001 \\
2\delta U^t & 0.0001 \leq \delta d_{\text{max}} < 0.01 \\
\delta U^t & 0.01 \leq \delta d_{\text{max}} \leq 0.05 \\
0.2\delta U^t & 0.05 < \delta d_{\text{max}} \leq 0.15 \\
0.1\delta U^t & \delta d_{\text{max}} > 0.15 \\
0.1Ue & \delta U^t \geq 0.1Ue \\
-10^{-9} & \delta U^t \leq 10^{-9}
\end{cases}
$$
Figure 16: Comparison of $\delta d_{\text{max}}$ between two displacement increment strategies

For $\delta U^t < 0$

$$
\delta U^{t+1} = \begin{cases} 
0.1\delta U^t & \delta d_{\text{max}} < 0.0001 \\
0.5\delta U^t & 0.0001 \leq \delta d_{\text{max}} < 0.01 \\
\delta U^t & 0.01 \leq \delta d_{\text{max}} \leq 0.05 \\
5\delta U^t & 0.05 < \delta d_{\text{max}} \leq 0.15 \\
10\delta U^t & \delta d_{\text{max}} > 0.15 \\
-0.1Ue & \delta U^t \leq 0.1Ue \\
10^{-9} & \delta U^t \geq -10^{-9}
\end{cases}
$$

with $\delta U^1 = 0.1Ue$ at $t=1$. $Ue = 0.064$ is obtained at $t = 0$. The $Z$ direction external force as a function of the prescribed displacement is presented in Figure 18. A snap-back behavior can be found.

Figure 19 illustrates the crack in cast iron at $t=100$. Different from the crack propagation in homogeneous materials, the crack path is strongly affected by the graphite nodules. As obtained experimentally under cyclic loading in [2], the presence of graphite nodules also strongly affected the crack initiation and propagation. The crack surface is therefore no-longer a plane as presented in Figure 20. Meanwhile, multi-cracking can be found. It confirms the ability to
6 Conclusions and perspectives

6.1 Conclusions

In this work, a new and efficient numerical strategy is proposed to solve phase field modeling of fracture problems. The matrix-free algorithm is applied to reduce memory requirement and improve parallel performance. PCGMG algorithms permit one to handle large jumps due to the presence of crack. The automatic load control strategy allows us to observe the snap-back behavior during the crack propagation. Several complicated simulations of fracture problems in heterogeneous materials are demonstrated for the first time. The numerical results confirm several phenomena observed during experiments. The micro-macro interactions are well revealed by these complicated numerical examples. It confirms the effect of the microstructure on the macroscopic behavior. This work opens up a new path for solving large scale 3D fracture problems.

6.2 Perspectives

The automatic load control strategy proposed in this work is not perfect. It should be adapted for different problems. A more general strategy could be investigated in the future. However, the
proposed strategy ensures that the local dissipation does not exceed a given threshold related to the maximum value of the rate of $d$. In the obtained load-displacement curves, it is observed that our strategy leads sometimes to unrealistic elastic unloading-reloading due to a too high decrease of the displacement increment. But, as there is no spurious energy dissipation during these artifacts, the envelop of the load-displacement curve as well as the damage evolution over time is well controlled and not biased.

Acknowledgments

The support of Région Pays de la Loire and Nantes Métropole through a grant Connect Talent IDS is gratefully acknowledged.

A A short comparison between the automatic load-control and the arc length strategy

To validate the proposed load control strategy, a short comparison is proposed using a crack nucleation problem for a one-dimensional bar with a reduced cross section under tension. The
Figure 19: Crack initialization in cast iron. Graphite nodules in green and the crack in red

reference example is taken from [35]. The geometry of this 1D bar is presented in Figure 21. The same parameter is applied according to [35]. Its length $L$ equals to 1 mm. The cross section $A$ is set to 1 mm$^2$. This bar is composed of a linear-elastic material with $E = 10$ MPa and $g_c = 0.1$ N/mm. The origin of axis X is at the center of the bar. It holds therefore a domain with $[-L/2, L/2]$. At $x = -L/2$, the bar is fixed. A prescribed displacement $Ut$ controlled using the proposed strategy is applied at $x = L/2$. The internal length scale $\ell_c$ is set to 0.05$L$. The element size $h$ equals to $\ell_c/32$.

The following automatic load control system is applied: for $\delta U_t > 0$

$$
\delta U_{t+1} = \begin{cases} 
2\delta U_t & \delta d_{\max} < 0.001 \\
\delta U_t & 0.001 \leq \delta d_{\max} \leq 0.00125 \\
0.001\delta U_t & \delta d_{\max} > 0.00125 \\
10^{-3} & \delta U_t > 10^{-3} \\
-10^{-4} & \delta U_t < 10^{-5}
\end{cases}
$$
Figure 20: Crack surface in cast iron at $t = 2390$

Figure 21: Bar of length $L$ with a reduced cross section in the center

for $\delta U^t < 0$

$$
\delta U^{t+1} = \begin{cases} 
0.001\delta U^t & \delta d_{max} < 0.001 \\
\delta U^t & 0.001 \leq \delta d_{max} \leq 0.00125 \\
2\delta U^t & \delta d_{max} > 0.00125 \\
-10^{-3} & 2\delta U^t < -10^{-3} \\
10^{-5} & \delta U^t > -10^{-5} 
\end{cases}
$$
with $\delta U^t = 10^{-3}$ at the first time step.

The distribution of $d$ along the bar at $d_{\text{max}} = 0.99$ is presented in Figure 22. The load–displacement curve is also obtained (see Figure 23). These two figures show that the result obtained by our strategy is similar to [35]. A small difference for the maximum load can be marked and also for the shape of $d$. However, note that different crack density functions are used in this paper and their work.

Figure 22: Distribution of the phase field parameter $d$ along the bar at $d_{\text{max}} = 0.99$

Data availability

The raw/processed data required to reproduce these findings cannot be shared at this time as the data also forms part of an ongoing study.
Figure 23: Load – displacement curves for $d_{max} = 0.99$

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