Evolution of interfaces, triple points, and quadruple points using discrete interface equations of motion

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Abstract
The evolution of interfaces is intrinsic to many physical processes ranging from cavitation in fluids to recrystallization in solids. Computational modeling of interface motion entails a number of challenges, largely due to the range of topological transitions that can occur over the course of the simulation. Microstructure evolution in a polycrystalline material that involves grain boundary motion is a particularly complex example due to the extreme variety, heterogeneity, and anisotropy of grain boundary properties. Accurately modeling this process is essential to determining processing-structure-property relationships in polycrystalline materials though, making modeling efforts central to the eventual rational design of materials. Simulations of microstructure evolution often use diffuse interface methods, and specifically the phase field method; these are advantageous for their versatility and ease of handling complex geometries, but can be prohibitively expensive due to the need for high interface resolution. Discrete interface methods require fewer grid points and can consequently exhibit better performance but have received comparatively little attention, perhaps due to the difficulty of consistently implementing topological transitions. This work considers a recently developed discrete-interface method for microstructure evolution, and provides an explicit comparison to diffuse interface methods by comparing against a multiphase field method in a number of classical microstructure evolution problems: the shrinking of a spherical grain, determination of a triple junction dihedral angle, and determination of a quadruple point dihedral angle. In each case, the discrete method is found to meet or outperform the multiphase field method in terms of accuracy for comparable levels of refinement, demonstrating its potential efficacy as a numerical approach for microstructure evolution.

Keywords: Discrete interface methods, diffuse interface methods, finite element analysis, phase field method, microstructure evolution

1. Introduction
The simulation of physical systems often requires the modeling of moving interfaces. This includes both interfaces at the boundaries between phases of matter including liquid/gas (e.g. cavitation), gas/solid (e.g. deflagration, sublimation, deposition), and solid/liquid (e.g. melting, solidification), and within a single solid phase. The grain structure of polycrystalline
materials in particular contains an extensive network of interfaces known as grain boundaries between regions of different crystallographic orientation. The interfacial dynamics governing this grain structure are often complex, leading to topological transitions and rapidly changing grain morphology that pose a unique type of modeling challenge.

A material’s grain structure is essential to its macroscopic properties. For example, solute segregation to grain boundaries can change the grain boundary cohesive energy \([11, 2]\), or the material’s susceptibility to hydrogen embrittlement \([3, 4]\) and stress corrosion cracking \([5, 6]\). Grain boundaries can provide preferential sites for the precipitation of a second phase, increasing or decreasing the plasticity of the polycrystal \([7, 8]\). They provide obstructions to the propagation of slip, with implications for the strength of the material as evidenced by the Hall-Petch equation \([9, 10]\). Given such consequences of the grain structure, it is not surprising that a variety of methods to simulate grain boundary motion and the evolution of the grain structure have been proposed in the literature \([11–16]\). Most of these represent the grain boundaries implicitly, as the locus of points where an indicator function abruptly changes value; such methods are referred to as diffuse interface methods below. While this has the advantage of not requiring that changes to the grain boundary topology be handled explicitly, the implicit representation complicates simulations of grain boundaries whose properties depend on their crystallography. This is not an issue when investigating the properties of generic grain boundary networks since a microstructure with constant and isotropic grain boundary properties is the canonical model system \([17]\). It does limit the possibility of predicting the properties of physical materials though, and therefore is a significant obstacle to realizing the vision of integrated computational materials design.

There have been comparatively few attempts to simulate the evolution of the grain structure explicitly \([18–21]\); such methods are referred to as discrete boundary methods below. The reason for this seems to be a practical one, namely, the difficulty with maintaining a consistent mesh of the grain boundary surfaces during a topological transition. Indeed, there did not even appear to be a way to enumerate a broad class of possible topological transitions in a general grain boundary network until quite recently \([22]\), and it is not at all obvious how to explicitly implement such transitions without at least knowing what they are. Nevertheless, discrete boundary methods do offer several distinct advantages with respect to microstructure modeling. They often require far fewer mesh points than their diffuse counterparts, offering a dramatic reduction in runtime computational cost. They also allow various defect properties, including grain boundary energies and triple line energies, to be explicitly defined in a way that is difficult with diffuse boundary methods. Along these lines, Kuprat previously developed GRAIN3D to simulate grain growth \([19]\) on a volumetric mesh with the gradient weighted moving finite element (GWFE) method, though the proposed topological transitions were not necessarily physical and could substantially affect the microstructure trajectory. Shya and Weygand instead proposed a method to simulate grain growth on a surface mesh and handled topological transitions by decomposing them into sequences of elementary operations \([20]\), but did not offer any assurance that such decompositions would not change the microstructure trajectory. Lazar et al. \([21]\) proposed a discretized formulation of the MacPherson and Srolovitz relation \([23]\) to simulate ideal grain growth on a surface mesh. While this only required that a small number of topological transitions be implemented, the explicit assumption of isotropic grain boundary properties precludes simulations of more general systems.
Two of the authors recently proposed a discrete interface method that addresses several of the computational challenges associated with explicit microstructure meshing, including a way to construct all possible topological transitions around a junction point and an energetic criterion to select one to apply [22]. The implementation [24] is based on SCOREC [25] and uses a volumetric microstructure mesh, potentially allowing the addition of other necessary physics to build a general framework for realistic simulations of microstructure evolution. SCOREC is an open source, massively parallelizable finite element framework with the adaptive meshing capabilities that are necessary to reach representative material volumes and to efficiently maintain the mesh element quality and desired degree of refinement. This work offers an initial comparison between this discrete interface method and a more well-established phase field approach.

This paper specifically introduces a set of three test cases to evaluate the relative accuracy and numerical cost of simulations of grain boundary motion, and uses this set to compare the discrete interface and phase field methods. The three cases correspond to several of the simplest configurations involving the motion of a grain boundary surface, a triple junction (TJ), and a quadruple point (QP). The grain boundary properties are assumed to be isotropic; a coarsening grain structure with isotropic grain boundary properties is said to be the ideal grain growth system, and provides a basis for comparison of all other evolving grain structures. Analytical forms for the evolving geometries are known for the spherical surfaces and TJ cases [26, 27], and the TJ and QP configurations have well-defined steady-state geometries. It is also of interest whether the two methods converge to the same geometries in situations for which analytical solutions are not known, since there is likely no other way to verify the simulations in such cases. While several of these configurations have been studied before, they are not usually considered in conjunction despite the benefits of doing so. Namely, the increasing complexity of the grain boundary configurations among the three test cases introduce different sources of systematic error to the grain boundary motion, and these errors can be more easily identified by comparing the test cases to one another.

It is desirable to establish the natures of any systematic errors and the accuracy of the simulation methods for a system with isotropic grain boundary properties before attempting to do so with more general grain boundary energy and mobility functions. The two methods considered in the present work will be capable of simulating the motion of grain boundaries with anisotropic properties when such functions become available. The discrete boundary method uses equations of motion that allow for general grain boundary properties and grain boundary lines that join an arbitrary number of grains [28]. The multiphase field model was developed to simulate the faceting of grain boundaries with energies that depend on boundary plane orientation, though this requires calculating a fourth-order derivative of the order parameters [29].

This paper is structured as follows. We begin with a discussion of the discrete interface method and its implementation, followed by an analogous discussion for the phase field/diffuse interface counterpart. We then apply both methods to a set of three test cases: a two grain system (shrinking sphere), a three grain system (triple junction), and a four grain system (quadruple point). The behavior of the discrete and diffuse models is compared for each of the examples vis-à-vis analytic predictions and the models’ internal length scales. The performance of the discrete interface case is briefly discussed, and then we conclude with a general discussion of the behaviors of the two models and a summary of recommendations.
2. Methods

Assuming that grain boundary properties are independent of grain boundary crystallography implies that the grain boundary network evolves along the negative gradient of the total boundary area. This is usually expressed as the Turnbull equation \[ v = m \gamma K \hat{n}, \] (1) governing the motion of each boundary patch where \( v \) is the velocity, \( m \) and \( \gamma \) are the mobility and energy per unit area, \( K \) is the mean curvature (the sum of the principle curvatures), and \( \hat{n} \) is the unit normal vector.

While this is sufficient to determine the time evolution of a closed surface, the Turnbull equation does not specify what happens at the TJs or QPs of the grain boundary network. One of the essential differences between discrete and diffuse interface models is the governing equations for precisely these locations. Discrete interface models generally represent the TJs and QPs as distinct entities with explicit geometries, and sometimes provide additional governing equations specific to these locations [21]. This is in contrast to the implicit approach of most diffuse interface methods which do not track TJs or QPs explicitly (while some diffuse interface methods do include higher order terms to account for the distinct behavior of line or point defects, these can come at extreme computational cost). In diffuse interface methods, each surface evolves according to the Turnbull equation with the geometric singularities at the TJs and QPs regularized by the diffuse interfaces. This difference in the handling of TJs is significant since the TJs define the geometric conditions at a grain boundary’s edges and thereby constrain the evolution of the grain boundary surface. Therefore, the microstructure trajectory is likely sensitive to the geometry around and motion of the TJs. It is for this reason that the angles between adjoining grain boundary surfaces are often used as simple scalar measures of the simulation accuracy in Section 3 below.

2.1. Discrete interface model

As implied by the name, every discrete interface model uses a discrete representation of the grain boundary network. A discrete representation entails that the grain boundary network geometry is represented by a collection of simple geometric objects, or elements, along with a description of how to join those elements together. The result is known as a surface mesh in three dimensions, and can be advantageously extended to a volumetric mesh to provide a discrete representation of the grain interiors as well. VDlib [22, 24] is a C++ library based on SCOREC [25] that represents a grain structure by means of a volumetric mesh containing tetrahedra, triangles, edges, and vertices.

There are two operations involved in updating the mesh to evolve the microstructure. The first moves the vertices of the mesh according to established equations of motion [28] that allow for anisotropic surface energies and arbitrary drag coefficients (the counterpart to the usual grain boundary mobility). The idea is that the velocity \( v \) of any given vertex should be such that the driving force \( F \) on the vertex is precisely balanced by the sum of drag forces \( Dv \) resulting from the motion of the adjoining grain boundary elements, where
Figure 1: Vectors describing the geometry around a vertex of the surface mesh. The central vertex is connected to five edges \( t_i \) and five triangles with unit normal vectors \( \hat{n}_{ij} \). The TJ along the edges \( t_1 \) and \( t_4 \) is shown in bold.

\( D \) is the drag tensor and \( v \) is the grain boundary velocity. The capillary force acting on the vertex is given by

\[
F = \sum_i \hat{t}_i \tau_l(\hat{t}_i) + \frac{1}{2} ||t_i|| \sum_{j:(i,j) \in \Delta} (\hat{n}_{ij} \times \hat{t}_i) \gamma(\hat{n}_{ij}) + \hat{n}_{ij} \frac{\partial \gamma}{\partial \phi_i} |\hat{n}_{ij}|, \tag{2}
\]

where \( \tau_l \) and \( \gamma \) are the line and surface energy functions, \( t_i \) the vector along edge \( i \) starting at the vertex and \( \hat{t}_i \) is the corresponding unit vector, \( \hat{n}_{ij} \) is the normal of the triangle formed by edges \( i \) and \( j \): \( \{i,j\} \in \Delta \) indicates an edge \( j \) starting at the vertex such that edges \( i \) and \( j \) span a triangle \( \Delta \), and \( \phi \) defines the surface orientation around edge \( i \); Fig. 1 shows several of these quantities for a generic vertex of a surface mesh. At force equilibrium the capillary forces are balanced by the drag forces \( Dv \) applied by the moving boundaries with

\[
D = \delta_0 I + \frac{1}{2} \sum_i \delta_1(\hat{t}_i) ||t_i|| (I - \hat{t}_i \otimes \hat{t}_i) + \frac{1}{6} \sum_{i,j \in \Delta} \delta_2(\hat{n}_{ij}) ||t_i \times t_j|| (\hat{n}_{ij} \otimes \hat{n}_{ij}) \tag{3}
\]

and where \( \delta_k \) is the drag term associated with the \( k \)-dimensional simplicial boundary element. The resulting boundary vertex velocity \( v \) is given by

\[
v = D^{-1} F. \tag{4}
\]

One advantage of this formulation is that the motion of every boundary vertex is governed by the same equation, including those on the interiors of surfaces, along TJs, and at QPs. If the point and line drag terms are zero, \( Dv \) reduces to the sum of the drag forces exerted by the neighboring triangles along the triangle normal directions for a given velocity \( v \). Moreover, if the grain boundary properties are constant, then \( \delta_2 = 3/m \) and this further reduces to a discrete version of Eq. (1) with an accuracy that depends on the product of the edge length and the mean curvature of the surface. The explicit nondimensionalized forms are given in Appendix A.

Apart from the motion of the mesh vertices, the accuracy of the discrete interface model is highly dependent on the element quality, where low-quality elements do not resemble equilateral triangles or tetrahedra [31, 32]. Without regular intervention and adaptation of
the mesh, the quality of mesh elements generically degrades with grain boundary motion, even to the point of elements inverting. The discrete interface method handles this by using MeshAdapt [33] to locally remesh where the element quality falls below a threshold value, and coarsening or refining edges with lengths below or above a threshold value. The target edge length \( \ell_e \) is constant in time and space for any given simulation, and an edge is coarsened or refined if the edge length \( l \) is outside the interval \( 0.7 \ell_e \leq l \leq 1.5 \ell_e \). These operations are used sparingly though, since apart from the computational expense local remeshing can perturb the grain boundary geometry. Specifically, these operations are the source of the discontinuous jumps observed in the discrete interface model results in Section 3.

2.2. Diffuse interface model

Comparison to a standardized diffuse interface model provides verification of the discrete interface model, as well as a benchmark for performance and code efficiency. In this work we apply the multiphase field model implemented following the presentation in Refs. [34, 35], which are general references for this section. A brief overview is provided here. For a system in region \( \Omega \subset \mathbb{R}^3 \) with \( N \) grains, \( N \) order parameters (denoted as the vector of functions \( \eta = \{\eta_1, \ldots, \eta_N\} \subset C_2(\Omega) \)) are defined such that the region occupied by grain \( i \) is precisely the support of \( \eta_i \). The free energy of the system is then defined to be

\[
W[\eta] = \int_{\Omega} \left( w(\eta) + \frac{1}{2} \sum_n k |\nabla \eta_n|^2 \right) dx,
\]

where \( w \) is the chemical potential and \( k \) is a model parameter to be discussed subsequently (the use of functional brackets should be understood to indicate dependence on the argument and any temporal or spatial derivatives). The following polynomial form is used for the chemical potential:

\[
w(\eta) = \mu \sum_n \left( \frac{1}{4} \eta_n^4 - \frac{1}{2} \eta_n^2 + \frac{3}{4} \sum_{m>n} \eta_m^2 \eta_n^2 \right), \quad \mu = 3.26.
\]

The coefficient for the boundary term is related to the grain boundary energy \( \sigma \) by

\[
k = \frac{3 \ell_{GB}}{4 \sigma},
\]

where \( \ell_{GB} \) is the diffuse boundary width. The evolution of \( \eta \), which determines the overall evolution of the microstructure, follows an \( L^2 \) gradient descent to minimize Eq. (5). The resulting kinetic evolution equation, expressed in terms of the variational derivative, is

\[
\frac{\partial \eta_n}{\partial t} = -L \frac{\delta W}{\delta \eta_n},
\]

where the rate coefficient \( L \) is related to the traditional boundary mobility \( m \) by

\[
L = \frac{4}{3} \frac{m}{\ell_{GB}}.
\]
All diffuse boundary calculations are performed using Alamo, a high performance multi-physics code that uses block-structured adaptive mesh refinement (BSAMR) with a strong-form elasticity solver to perform diffuse interface calculations \[36\]. Alamo is built on the AMReX package, developed by Lawrence Berkeley National Laboratory \[37\]. All of the results presented here were run on a desktop computer and generally completed in less than an hour (depending on the chosen parameters). Of particular interest is the convergence of the solution with respect to the boundary width, \(\ell_{GB}\), which determines the diffuse boundary length scale. The exact solution is recovered as \(\ell_{GB} \to 0\), but this comes at the expense of increased computational cost. In this work we are particularly interested in the relationship between \(\ell_{GB}\) and the discrete interface model counterpart.

### 3. Results and discussion

Three cases are considered in this section to quantify the systematic error of the discrete interface and phase field methods of simulating grain boundary motion. The first is a spherical grain which evolves in a self-similar way. This is a standard configuration that is often used in the literature to verify that the Turnbull equation is obeyed in the absence of complicating factors \[16, 19, 38, 39\]. The second is a TJ that migrates along a semi-infinite grain boundary \[16, 40, 41\], eventually reaching a steady state configuration with a known profile and velocity \[26\]. The quantity considered below is the angle of the grain boundaries at the TJ, though in principle a stricter validation scheme could involve evaluating the simulation’s ability to precisely reproduce the expected geometric evolution. The third is a columnar hexagonal grain configuration that migrates along semi-infinite grain boundaries to allow a study of the steady state evolution of a QP \[16, 42\]. Perhaps the reason this case appears less often in the literature is that an analytical solution for the boundary profile is not known; instead, the angles between grain boundary traces on two cross-sections are evaluated for convergence and used to compare the two simulation methods. The grain boundary geometries for the three test cases are described in their respective sections, have Neumann boundary conditions, and are constructed to make the grain boundary curvatures comparable.

It is expected that the accuracy of both the discrete and diffuse interface models will increase with decreasing internal length scale \(\ell\), denoted as \(\ell = \ell_e\) for the discrete model and \(\ell = \ell_{GB}\) for the diffuse. However, the accuracy should not depend on any absolute length scale, since then the accuracy could be improved simply by uniformly scaling the grain structure. The accuracy therefore depends on \(\ell\) relative to a second length scale that is characteristic of the evolving interface. Since the accuracy should be invariant to the isometries of Euclidean space, the inverse of the interface’s mean curvature is the natural candidate for the second length scale, and the accuracy of both models is expected to depend on the dimensionless product of \(\ell\) and interface’s mean curvature. More precisely, all of the errors reported in this section are expected to be power laws in \(\ell\), with the prefactor depending on the mean curvature and implementation details in a way that is difficult to parameterize (only the spherical grain has the same mean curvature everywhere). For this reason, only the exponent of \(\ell\) is generally reported in the following.

Many of the quantities reported below are nondimensionalized following the procedure in Appendix A to facilitate the comparison of the discrete interface and phase field methods. A tilde indicates a nondimensionalized variable (with the exception of \(\ell_e\) and \(\ell_{GB}\) which
are always nondimensionalized) and an analytical prediction is denoted by the subscript \( t \), e.g., \( \tilde{r}_t(\hat{t}) \) is the analytical prediction for the nondimensionalized radius of the sphere as a function of nondimensionalized time. The equations of motion of the discrete interface method were integrated using a second order Runge–Kutta scheme with a maximum time step of \( 1.2500 \times 10^{-5} \).

3.1. Spherical grain

The spherical grain case is intended to reveal the error when modeling surface motion in the absence of confounding effects from other grain boundary network components. One advantage of this particular choice is that, provided the grain boundary properties are constant and isotropic, the evolution of a spherical grain is known analytically. As derived in Appendix B, the sphere shrinks uniformly with radius

\[
\tilde{r}_t(t) = \sqrt{r_0^2 - 4m\gamma t} \tag{10}
\]

as a function of time. Nondimensionalizing this equation reveals that a sphere starting with a radius of \( \tilde{r}_t(\hat{t}) = 1 \) vanishes at \( \hat{t} = 0.25 \). The actual simulations deviate from Eq. (10) both because the initial geometries shown in Fig. 2 are not precisely spheres and because the Turnbull equation in Eq. (1) is not precisely followed, though these sources of error are reduced as the \( \ell \) are made smaller. Since the diffuse interface model does not perform well when the radius of the sphere approaches the grain boundary width, the magnitude of the error for the shrinking grain is quantified by the deviation of the sphere half-life \( t_{\text{half}} \) from the analytical prediction

\[
t_{\text{half},t} = \frac{3r_0^2}{16m\gamma}. \tag{11}
\]

When nondimensionalized, this reduces to \( \hat{t}_{\text{half},t} = 3/16 \).

Figure 3 shows the performance of the two models, with the discrete interface model on the left and the diffuse interface model on the right. The top row shows the radius of the sphere as a function of time, where the color indicates the internal length scale and the exact solution is in black. The roughness of the curves for the discrete interface model is due to remeshing to preserve the element quality, and the velocity in the diffuse interface model falls as the radius approaches the grain boundary width. The magnitude of the relative error
in the radius as a function of time is shown in the middle row. The error for the discrete interface model is caused by the magnitudes of the surface vertex velocities being larger than predicted by the analytical solution, perhaps as a consequence of the equations of motion being explicit and uncoupled. That the accumulation of error accelerates with decreasing radius supports the hypothesis that the error generally depends on the product of $\ell_{e}$ and the mean curvature. Meanwhile, there are likely two sources of error that contribute to the results for the diffuse interface model. The error at early times is a postprocessing artifact that occurs when constructing isocontours to identify the location of the grain boundary, effectively resulting in an offset to the sphere radius. The other source of error relates to the order parameter gradient at a grain boundary patch being affected by the presence of nearby patches. This is most visible when the grain is about to collapse and grain boundary patches on opposite sides of the grain interact, reducing the gradient magnitude and the grain boundary velocity. Conversely, the mean curvature of the surface causes neighboring grain boundary patches to interact, increasing the gradient magnitude and the grain boundary velocity at earlier times. As with the discrete interface model, the magnitude of this effect at
earlier times is proportional to the product of $\ell_{GB}$ and the mean curvature.

The bottom row of Fig. 3 shows the half-life error $|\tilde{t}_{\text{half}} - \tilde{t}_{\text{half},t}|$ as a function of the internal length scale. A conjugate gradient minimization algorithm and bootstrapping were used to fit $|\tilde{t}_{\text{half}} - \tilde{t}_{\text{half},t}|$ to a power law in the internal length scale $\ell$. This gives an exponent of $1.37 \pm 0.21$ for the discrete interface model and $0.678 \pm 0.085$ for the diffuse interface model, where the values are the medians and the uncertainties are half the interquartile range. While the exponents could suggest that the error of the diffuse interface model decays slower than that of the discrete interface model with decreasing internal length scale, the errors in the apparent grain radius due to isocontour construction during postprocessing do not actually affect the microstructure trajectory. This could motivate using the two-grain configuration with self-similar evolution analyzed by Mullins [26] in the future since such postprocessing errors would likely not affect the long-time behavior.

3.2. Triple junction

The purpose of the TJ case is to include a TJ in the moving boundary while keeping the grain configuration as simple as possible, ideally allowing the error of the equations of motion for the TJ to be identified by comparing the results to those for the spherical grain. The initial geometries of the grain configuration are shown in Fig. 4, are constant in the out-of-plane direction, and have mirror boundary conditions in the lateral directions. The rate of volume change of the top grain can be derived by applying the von Neumann-Mullins equation [26, 43] to the two-dimensional grain configuration in a plane perpendicular to the TJ. Since there is one triple point per simulation cell in this plane, the rate of cross-sectional area change of the top grain per simulation cell width $L$ is $\pi m \gamma / 3$, and the rate of volume change of the top grain can be found by multiplying by the TJ length. Mullins actually went further and solved for the steady-state profile of the moving boundary assuming constant and isotropic grain boundary properties [26]. If $x$ is distance from the left edge of the simulation cell and $y$ is height from the top of the red grain, then the steady-state profile of the grain boundary between the red and blue grains is

$$y(x) = -\ln[\cos(\pi x)]/\pi.$$  \hspace{1cm} (12)

The width of the simulation cell is defined by the above equation as $L = 1/3$ which can be scaled to match the dimensions of the simulation cell.
The dihedral angle $\theta_{TJ}$ between the two boundaries of the blue grain is perhaps the simplest way to evaluate the accuracy of the geometry of the moving boundary in the vicinity of the TJ. A force balance argument for constant and isotropic grain boundary properties (and the absence of any TJ drag) leads to the condition $\theta_{TJ,t} = 2\pi/3$. This is roughly enforced in the initial conditions by defining the two parts of the moving boundary to be the appropriate sections of of cylinders; while this is not the steady-state profile given by Mullins, it is sufficiently close for a short initial transient and rapid convergence to the steady-state condition as is visible in Fig. 5.

As before, results for the discrete interface model are on the left and those for the diffuse interface model are on the right. The top row shows $\theta_{TJ}$ as a function of time, where the color indicates the internal length scale and the exact solution $2\pi/3$ is in black. The roughness of the curves for the discrete interface model is due to the remeshing required to maintain element quality, and the periodic spikes that appear for the diffuse interface model are due to the interaction of the adaptive mesh refinement and the construction of the isocontours. The error in $\theta_{TJ}$ (measured as the median of the second half of the time series) is shown in the bottom row, with the dependence of the steady-state angle on $\ell_e$ for the discrete interface model being a consequence of the linear elements forcing the grain boundary curvature to be concentrated at the vertices and edges of the mesh. Specifically, the grain boundary curvature that is distributed to the TJ edges causes the deviation of $\theta_{TJ}$ from the expected value, with the magnitude of the deviation depending on the product of $\ell_e$ and the mean curvature of the adjoining grain boundary. Identifying the precise location of the TJ and the value of $\theta_{TJ}$ is more difficult for the diffuse interface model since the grain boundary geometry is implicit. The procedure followed here involves fitting third- and fourth-order polynomial approximations to each side of the isocontour where the order parameter for the top grain is 0.5. The triple point location in the plane is then defined to be the point of intersection of the polynomials, and $\theta_{TJ}$ is the angle between the tangent vectors at the point of intersection. This process works well in the sharp interface limit, but is very sensitive to perturbations in the solution for larger $\ell_{GB}$ since there is substantially more error in the predicted location of the TJ with respect to the simulation size. The occasional deviations that are observed in the steady-state correspond to BSAMR re-gridding events.

Fitting a power law in the internal length scale $\ell$ to $|\theta_{TJ} - \theta_{TJ,t}|/\pi$ gives an exponent of $0.91 \pm 0.20$ for the discrete interface model and $1.45 \pm 0.13$ for the diffuse interface model, where the values are the medians and the uncertainties are half the interquartile range. The additive offset of $-0.005 \pm 0.011$ for the discrete interface model is entirely consistent with $\theta_{TJ}$ converging to the equilibrium angle in the $\ell_e \to 0$ limit, though at a lower rate than the half-life error magnitude in Fig. 3. This is not unexpected though, since the TJ can be thought of as a jump condition in the tangent plane to the grain boundary that is both difficult to accurately reproduce with a finite element mesh and is not present in the spherical grain case. While the exponent for the diffuse case is nominally higher, this is not reflective of the trend observed for small $\ell_{GB}$ where the saturation in the error is likely the result of inaccuracy in the postprocess calculation of the angle. The higher exponent therefore does not necessarily indicate better convergence.
Figure 5: Comparison of $\theta_{TJ}$ for discrete model (left) and diffuse model (right); all quantities are nondimensionalized. (Top row) Plot of $\theta_{TJ}$ vs time, with color indicating the length scale and the exact solution in black. (Bottom row) Plot of the relative error vs length scale.

3.3. Quadruple point

As with the TJ case, the grain structure for the QP case consists of a top grain above several columnar grains. The grain boundaries of the top grain migrate down the simulation cell, consuming the columnar grains and eventually reaching a steady-state profile, though an analytical solution for this profile is not known. The configurations of columnar grains for the discrete and diffuse interface models are shown in Figs. 6a and 6b respectively, with the hexagonal cross-sections of the columnar grains clearest for the discrete interface model; the BSAMR mesh makes simulations of rectilinear domains like the one in the figure strongly preferable for the diffuse interface model.

Following the initial transient, the steady-state profile is examined on the two planes indicated in Fig. 6c, one along a minor diameter of the central grain and bisecting a TJ, the other along a major diameter of the central hexagonal grain and containing a QP. The angles along these profiles at the intersections with the TJ and the QP are reported in Fig. 7 and Fig. 8. While the equilibrium angle at the TJ should be $2\pi/3$ (the same as for the TJ in Section 3.2), the curvature of the grain boundaries in both principal directions could change the rate of convergence to $2\pi/3$ with decreasing $\ell$ compared to the TJ case. As for the equilibrium angle at the QP, an infinitesimal neighborhood of the QP will contain triple junction lines in a tetrahedral configuration connected by flat grain boundary surfaces provided the principal curvatures of the grain boundaries are finite. This allows the equilibrium angle of $\cos^{-1}(-1/\sqrt{3}) \approx 0.696\pi$ at the QP along the major diameter to found by geometrical considerations.

Starting with the TJ angle, observe that the data points for the TJ angle error along the minor axis in the bottom row of Fig. 7 closely resemble those for the TJ angle error in
Figure 6: QP mesh configurations and schematic. (a) Hexagonal columnar grain mesh for the discrete interface model. (b) Hexagonal columnar grain BSAMR mesh in a rectilinear domain for the diffuse interface model. (c) Locations of QP and TJ along major and minor lines.

the bottom row of Fig. 5. This indicates that the nonzero second principal curvature of the grain boundaries along the TJ lines in the QP case does not have a significant effect on the error in the equations of motion, and is consistent with the expectation that the error should scale with the mean curvature (the sum of the principal curvatures). Fitting a power law in the internal length scale $\ell$ to $|\theta_{TJ} - \theta_{TJ,t}|/\pi$ gives an exponent of $0.927 \pm 0.223$ for the discrete interface model and $0.85 \pm 0.50$ for the diffuse interface model, with both models converging to the expected value. While the exponent for the discrete interface model is nearly identical to that for the TJ case, the lower exponent for the diffuse interface model is likely a consequence of a power law fitting the data relatively poorly; observe that the TJ angle error for the diffuse interface model does not fall on a line on a log-log plot, and instead seems to saturate at a lower bound set by the angle estimation procedure in postprocessing.

For the QP angle, the final values for the discrete interface model follow a power law in $\ell$ that converges to an angle of $(0.694 \pm 0.001)\pi$ with an exponent of $0.958 \pm 0.026$, whereas the respective values for the diffuse interface model are $(0.707 \pm 0.011)\pi$ and $0.85 \pm 0.45$; the limiting values for both the discrete and diffuse interface models effectively coincide with the exact value. It is significant that the errors for all of the discrete interface results in Secs. 3.2 and 3.3 decay with exponents that are close to one. The discrete interface method uses linear elements that approximate the grain boundary geometry with first-order accuracy, meaning that an exponent of one is the best possible result. It is likely that higher-order elements would need to be used to substantially increase the rate of error decay with $\ell_e$. The irregularity in the exponents for the diffuse interface model in Secs. 3.2 and 3.3 is attributed to the error in the polynomial algorithm used to extract the grain boundary profile. Examination of Figs. 5 and 7 indicates that this functions as a source of random error that is larger for highly diffuse boundaries but vanishes in the sharp boundary limit.

4. Performance

When selecting a numerical method in practice, computational cost is often nearly as much a concern as the accuracy of the simulated behavior. This section specifically considers the dependence of the discrete interface method’s computational cost on the internal length scale $\ell_e$. Suppose that the main contribution to the computational cost is evaluating the equations of motion for the grain boundary vertices. The number of such vertices is expected to depend on the internal length scale as $\ell_e^{-2}$. If the velocity of the vertices is independent of
Figure 7: Comparison of minor axis results for the QP case for the discrete model (left) and diffuse model (right); all quantities are nondimensionalized. (Top row) Plot of the measured TJ (minor diameter) angle, with color indicating the length scale and the exact solution in black. (Bottom row) Plot of the relative error in the TJ angle with respect to length scale.

Figure 8: Comparison of major axis results for the QP case for the discrete model (left) and diffuse model (right); all quantities are nondimensionalized. (Top row) Plot of the measured QP (major diameter) angle, with color indicating the length scale and the exact solution in black. (Bottom row) Plot of the relative error in the QP angle with respect to length scale.
Discrete edge length ($\ell_e$), then the time step length should decrease as $\ell_e$ to keep the vertex displacement shorter than the characteristic edge length and prevent mesh element inversion. This would imply that the overall computational cost should scale with $\ell_e^{-3}$, or as the product of the number of grain boundary mesh vertices and the number of time steps for a given overall simulation time.

Figure 9 shows the scaling of the normalized runtime cost and normalized number of grain boundary vertex calculations $n_{\text{calc}} = \sum_j n_{\text{vb},j}$, where $n_{\text{vb},j}$ is the number of grain boundary vertices at time step $j$, for the discrete interface method. These scale as $\ell_e^{-4.088}$ and $\ell_e^{-4.081}$, respectively, for small $\ell_e$ where the computational cost of the vertex calculations is expected to dominate. This confirms that the overhead of the discrete interface method (mesh management, enumeration of topological transitions, etc.) is relatively small compared to the evaluation of the equations of motion, but is not consistent with the $\ell_e^{-3}$ scaling expected in the previous paragraph. This discrepancy is a result of the length of the median time step scaling as $\ell_e^{1.916}$ instead of linearly; the underlying cause for this time step scaling is investigated further in Appendix C.

5. Conclusion

The purpose of this work has been to establish the validity and performance of a recently-developed discrete interface method by comparison to analytic solutions and a well-established multiphase field method. More specifically, the evolution of the simplest configurations involving surfaces, triple lines, and quadruple points with self-similar behavior given constant and isotropic grain boundary properties are used to quantify the error in position and junction angles as a function of the degree of refinement. The boundary types are simple enough to be amenable to analysis, yet complex enough to introduce different systematic errors over the course of their evolution. Despite the approaches for simulating boundary motion being distinctly different, our results indicate that both methods converge to the same junction angles with similar rates. The most significant difference is that when predicting the half life of the shrinking sphere, the convergence rate of the diffuse interface method appears to be about half of that of the discrete interface method.
Although this work assumes constant and isotropic grain boundary properties, both methods were developed with the intention of performing simulations for anisotropic grain boundary properties. There are reasons to believe that the necessary data could be available in the future. Morawiec [44] suggested that the grain boundary energy could be experimentally obtained as a function of the grain boundary crystallography by applying the Herring condition [45, 46] to triple junctions imaged by three-dimensional microscopy techniques [47, 48]. Alternatively, molecular dynamics simulations allow direct evaluation of grain boundary properties in bicrystals [49]. While the excessive number of points required to adequately sample the five-dimensional grain boundary space of has precluded the availability of general grain boundary energy and mobility functions in the literature, there has been progress for particular subsets of grain boundaries [50, 51]. The absence of general grain boundary energy and mobility functions has practically limited existing simulations of grain boundary motion to the canonical system for which the grain boundary properties are assumed to be independent of grain boundary crystallography.

The performance of the discrete interface model lends confidence in its ability to yield accurate results for more general and complex microstructures for which there is no known analytic solution. Moreover, the performance with unoptimized code indicates reasonable scaling behavior that is close to the ideal scaling and comparable to that of alternative methods.

Acknowledgements

EE and JKM were supported by the National Science Foundation under Grant No. DMR 1839370. BR was supported by the Office of Naval Research, grant #N00014-21-1-2113. This work used the INCLINE cluster at the University of Colorado Colorado Springs. INCLINE is supported by the National Science Foundation, grant #2017917.

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Appendix A. Nondimensionalization

Define the variable $L$ to be the characteristic length scale of the grain structure defined in Section 3. For a sphere it is the sphere radius, for the TJ it is the length of the simulation cell in the direction normal to the consumed grain boundary, and for the QP it is the hexagonal grain’s minor diameter. The Turnbull equation in Eq. (11) suggests that there is a characteristic time scale $\tau = L^2/(m\gamma)$. The simulations are performed with nondimensionalized time $\tilde{t} = t/\tau$, nondimensionalized space $\tilde{x} = x/L$, nondimensionalized rate of volume change $d\tilde{V}/d\tilde{t} = (\tau/L^3)dV/dt$, etc.

With respect to the quantities defined in Section 2.1, suppose that $\tau(t_i) = 0$, $\delta_0 = 0$, $\delta_1(t_i) = 0$, and that $\gamma(n_{ij})$ and $\delta_2(n_{ij})$ are constants. The governing equations of the discrete interface model then reduce to:

$$F = \frac{\gamma}{2} \sum_i \left| \dot{t}_i \right| \sum_{j: \{i,j\} \in \Delta} \dot{n}_{ij} \times \dot{t}_i, \quad (A.1)$$
$$D = \frac{\delta_2}{6} \sum_{i,j \in \Delta} \left| \dot{t}_i \times \dot{t}_j \right| (\dot{n}_{ij} \otimes \dot{n}_{ij}), \quad (A.2)$$
$$v = D^{-1}F. \quad (A.3)$$

The nondimensionalized versions of these equations are

$$\tilde{F} = \frac{F}{L^2\gamma} = \frac{1}{2} \sum_i \left| \dot{\tilde{t}}_i \right| \sum_{j: \{i,j\} \in \Delta} \dot{\tilde{n}}_{ij} \times \dot{\tilde{t}}_i, \quad (A.4)$$
$$\tilde{D} = \frac{D}{L^2\delta_2} = \frac{1}{6} \sum_{i,j \in \Delta} \left| \dot{\tilde{t}}_i \times \dot{\tilde{t}}_j \right| (\dot{\tilde{n}}_{ij} \otimes \dot{\tilde{n}}_{ij}), \quad (A.5)$$
$$\tilde{v} = \frac{\tau}{L} v = \frac{\tau\gamma}{L^2\delta_2} \tilde{D}^{-1} \tilde{F}, \quad (A.6)$$
where $\delta_2 = 3/m$ when the triple line and quadruple point drags vanish; this can be derived by requiring that the limiting behavior of a small spherical cap coincides with the predictions of Eq. (1).

The corresponding nondimensionalization of the multiphase field governing equations in Sec. 2.2 yields

$$\frac{\partial \eta}{\partial \tilde{t}} = -\tau L \frac{\delta W}{\delta \eta_n} = \frac{\partial \tilde{w}}{\partial \eta_n} + k \tilde{\Delta} \eta_n, \quad \tilde{w} = w \tau L, \quad \tilde{\Delta} = \tau L k \Delta.$$ (A.7)

In this work, all multiphase field calculations are performed with dimensional values and then nondimensionalized for comparison to discrete interface simulations.

**Appendix B. Spherical grain**

If grain boundary properties are constant and isotropic, then the evolution of a spherical grain is self-similar and can be completely described by the radius $r_t(t)$ as a function of time. Since the mean curvature is $K = 2/r_t$, Eq. (1) implies that

$$\frac{dr_t}{dt} = -2m\gamma/r_t.$$ (B.1)

Setting $t_0 = 0$ and $r_t(0) = r_0$ and integrating gives

$$r_t(t) = \sqrt{r_0^2 - 4m\gamma t}.$$ (B.2)

as the solution to this differential equation. Since the characteristic length scale for a sphere is $r_0$, nondimensionalizing reduces this to

$$\tilde{r}_t(\tilde{t}) = \sqrt{1 - 4\tilde{t}}.$$ (B.3)

for the black curve in Fig. 3.

**Appendix C. Scaling analysis**

As described in Sec. 4, while the computational cost of the discrete interface method is expected to scale as $\ell^{-3}$, the actual scaling is instead $\ell^{-4.089}$. Closer investigation revealed that the time step could decrease or increase by multiple orders of magnitude depending on the presence of various local mesh configurations. The boundary triangles exert capillary forces only in the boundary plane, yet exert drag forces only in the out-of-plane direction. This allows vertices on nearly-flat grain boundary sections to experience arbitrarily large lateral velocities, slowing the simulation down as the time step is reduced to prevent element inversion. The discrete method simulations in Sec. 3.1 include an isotropic contribution $D_{I,d} = A_m^2/(md) I$ to the drag tensor such that $v = (D + D_{I,d})^{-1} F$, where $A_m^2$ is the mean triangle area over the whole simulation, $m$ is the mobility, and $d = 1000$ is a drag ratio. Decreasing the drag ratio reduces the lateral velocities, but also slows down the actual motion of the boundary and introduces a systematic error.

As an alternative, a contribution to the drag tensor that only acts in the in-plane directions could be constructed as follows. For simplicity, consider a closed disk of coplanar triangles
Figure C.10: The effect of different drag tensor correction terms on the resulting velocity. The capillary force is colored black and the velocities corresponding to different correction terms are differentiated by color. Each vector is scaled relative to the maximum magnitude among the velocities.

Figure C.11: The scaling of the median time step with $\ell_e$ for the three drag tensor correction terms.

around a vertex. Iterating over each grain boundary triangle $\Delta_{ij}$ adjacent to the central vertex, find the relative positions of the other vertices from the central vertex $p_i$ and $p_j$ and construct the outer product of the difference $p_i - p_j$ with itself. Let $\Lambda_{\text{max}}$ be the largest eigenvalue of the sum of the outer products, and define the matrix $C = \sum_{i,j \in \Delta} (p_i - p_j) \otimes (p_i - p_j)/\Lambda_{\text{max}}$. The anisotropic drag tensor contribution $D_{a,d} = A_m^2/(md)C$ by construction has no effect on the grain boundary motion in the plane normal direction. This should allow the lateral velocities of boundary vertices to be reduced while introducing less systematic error in the motion of non-planar boundaries than for an isotropic drag. An example triple junction mesh configuration is shown in Fig. C.10 to qualitatively demonstrate the effect of different drag tensor correction terms. Although the velocity associated with $D_{I,d}$ aligns with the force direction faster with increasing $d$, the velocity term in the vertical direction is also attenuated more compared to $D_{a,d}$.

The difference in the expected and the actual scaling of the cost can largely be attributed to the non-linear scaling of the median time step shown in Fig. C.11. $dt_{\text{med}}$ scales as $\ell_e^{2.010}$ for $D_{a,10}$ and $\ell_e^{4.916}$ for $D_{I,1000}$ and $D_{I,10}$. Overall, $D_{a,10}$ allows larger time steps and has a better accuracy, though the improvement is not significant.