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A Boltzmann-type kinetic model for misorientation distribution functions in two-dimensional fiber-texture polycrystalline grain growth

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ABSTRACT

For mathematical modeling of polycrystalline materials, it is critical to understand how the statistics of evolving grain boundary networks depend on the set of laws that govern the dynamics at a microscopic scale. Such rules describe the motion of grain boundaries and their junctions, as well as criteria for the corresponding topological transitions. Although these laws have been known for quite some time, their precise role in the development of the networks' mesoscopic properties is not sufficiently clear. A possible way to establish a connection between statistical features of the whole networks and fundamental rules for the evolution of individual grain boundaries is to conduct numerical experiments via generally recognized large-scale simulation approaches. Alternatively, it is possible to develop less computationally expensive techniques such as kinetic models based on partial differential equations, while preserving essential characteristics of grain boundary networks. The aim of this paper is to develop a novel kinetic modeling approach for estimating number- and length-weighted misorientation distribution functions in the particular technologically important case of polycrystalline thin films with fiber textures and with anisotropic grain boundary energy densities. Our approach is tested by comparing the numerical results, which it gives for two specific benchmark examples, with the corresponding largescale simulation results from the related literature. One auxiliary benchmark example is also considered. © 2016 Acta Materialia Inc. Published by Elsevier Ltd. All rights reserved.

1. Introduction

Polycrystalline materials such as metals and ceramics are comprised of single crystallites called grains. Interfaces between grains are called grain boundaries. Orientations, shapes, and arrangements of grains have a direct relationship to macroscopic properties of materials. For instance, the presence of grain boundaries decreases thermal and electrical conductivity affecting the performance of chips in microprocessors. Grain boundaries disrupt motion of dislocations through materials. Therefore, reducing crystallites' sizes is a common way to improve strength and fracture toughness of materials, as described by the wellknown Hall–Petch relationship [1].

Grains' and grain boundaries' configurations, or microstructures, are determined by a variety of factors such as history of deformations, phase transitions, heat treatment, etc. In this paper, we are interested in the process of microstructural relaxation known as grain growth or coarsening. This phenomenon refers to an increase in size of grains at a high temperature, occurring when recovery and recrystallization are complete, and the further reduction in the surface energy can be achieved only by decreasing the total area (in three-dimensional cases) or the total length (in two-dimensional cases) of grain boundaries.

For mathematical modeling of polycrystalline materials, it is critical to understand how the statistics of evolving grain boundary networks depend on the set of laws that govern the dynamics at a microscopic scale. Such rules describe the motion of grain boundaries and their junctions, as well as criteria for the corresponding topological transitions. Although these laws have been known for quite some time, their precise role in the development of the networks' mesoscopic properties is not sufficiently clear. A possible way to establish a connection between statistical features of the whole networks and fundamental rules for the evolution of individual grain boundaries is to conduct numerical experiments via generally recognized large-scale simulation approaches such as Monte Carlo





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techniques [2–8], phase field models [6,9–11], curvature-driven grain growth models [12–15], vertex models [16–18], diffusion generated motion models [19,20], level set methods [21–23], etc.

Alternatively, it is possible to develop less computationally expensive techniques such as kinetic models based on partial differential equations, while preserving essential characteristics of grain boundary networks.

It is necessary to mention the Fradkov mean-field type kinetic model for two-dimensional grain growth [29-32] which is based on the classical von Neumann–Mullins law and considers time-evolving distribution functions for grain areas and topological classes (i. e., numbers of sides per grain). This approach is applied only to isotropic systems, because the von Neumann–Mullins relation and even the corresponding triple junction dragging generalization [33, § 4.6] require that the grain boundary energy density is constant. As opposed to the Fradkov model and its possible generalizations, our goal is to estimate so-called number-and length-weighted misorientation distribution functions (as considered in Refs. [4–8,10,17,20,22]) for anisotropic systems.

A novel framework for kinetic modeling of anisotropic microstructure evolution with accompanying topological transitions was proposed in Ref. [24] and applied to a simplified one-dimensional model. A statistical mechanics perspective was adopted, and a system of grain boundaries was treated as a collection of interacting "particles", where the state of each particle was determined by parameters associated with the corresponding boundary. By exploiting grain boundaries' and particles' duality, two separate Boltzmanntype kinetic equations were derived, differing only in the number and choice of parameters constituting the corresponding state spaces.

The aim of this paper is to modify and extend the mentioned kinetic modeling approach to the two-dimensional case of polycrystalline thin films with so-called fiber textures [10]. In this case, grains have nearly identical orientations in the axial direction perpendicular to the considered film, but random radial orientations in the plane of the film [25]. Many technologically important characteristics of the films, such as their strength, conductivity, corrosion resistance or resistance against void formation, are correlated with fiber-texture characteristics of grain microstructures and grain boundary networks, in particular, with the misorientation distribution functions; see, for instance, [10,26–28].

Note that the model developed in this paper is not exactly kinetic, because instead of the time variable, it operates with the varying total number of grain boundaries which is assumed to be decreasing in time.

We test our approach by comparing the numerical results, which it gives for two specific benchmark examples, with the corresponding large-scale simulation results presented in Refs. [10,20]. One auxiliary benchmark example is also considered.

2. Statement of the problem

Consider a two-dimensional polycrystalline system with a perfect fiber texture [3,5,10,20]. This means the following:

- the grains and their boundaries have a common tilt axis (up to parallel transports) perpendicular to the plane of the simulation cell;
- there exists a single orientational degree of freedom for each of the grains, which is the azimuth (i. e., the angle of rotation) with respect to a fixed axis (up to parallel transports) in the plane of the simulation cell;
- during growth or shrinkage, the grains do not change their orientations.

In this context, the misorientation angle between two

neighboring grains characterizes the corresponding boundary.

Suppose that scalar grain orientations and scalar misorientations between two neighboring grains have the smallest positive period α_{max} , and let us identify every two opposite misorientations with equal absolute values. Then we get the misorientations' symmetry with respect to $\frac{\alpha_{max}}{2}$ on the interval $[0,\alpha_{max}]$, and the sets of all orientations and misorientations up to the whole crystal symmetry can be taken as $[0,\alpha_{max}]$ and $[0,\beta_{max}]$, respectively, where

$$\beta_{\max} \stackrel{\text{def}}{=} \frac{\alpha_{\max}}{2}.$$
 (1)

For instance, $\alpha_{\text{max}} = \frac{\pi}{2}$ and $\beta_{\text{max}} = \frac{\pi}{4}$ for the fourfold square symmetry, whereas $\alpha_{\text{max}} = \frac{2\pi}{3}$ and $\beta_{\text{max}} = \frac{\pi}{3}$ for the regular hexagon symmetry.

Let

$$\gamma: [0, \beta_{\max}] \to [0, +\infty) \tag{2}$$

be the grain boundary energy density depending only on the misorientation variable.

Let $t \in [0,T]$ be the time variable, and let the total number and length of grain boundaries in the considered system be given by N(t) and L(t), respectively. Also denote the total number and length of boundaries with a particular misorientation $\beta \in [0,\beta_{\text{max}}]$ by $N(\beta,t)$ and $L(\beta,t)$, respectively. Likewise, let $\overline{L}(t)$ be the average boundary length in the whole system, and let $\overline{L}(\beta, t)$ be the average length for a particular misorientation $\beta \in [0,\beta_{\text{max}}]$:

$$\bar{L}(t) \stackrel{\text{def}}{=} \frac{L(t)}{N(t)}, \quad \bar{L}(\beta, t) \stackrel{\text{def}}{=} \frac{L(\beta, t)}{N(\beta, t)}
\forall \beta \in [0, \beta_{\text{max}}] \quad \forall t \in [0, T].$$
(3)

Now introduce the number- and length-weighted misorientation distribution functions (MDFs) [8,10]:

$$\rho_n(\beta, t) \stackrel{\text{def}}{=} \frac{N(\beta, t)}{N(t)}, \quad \rho_l(\beta, t) \stackrel{\text{def}}{=} \frac{L(\beta, t)}{L(t)}$$
(4)

 $\forall \beta \in [0, \beta_{\max}] \quad \forall t \in [0, T].$

Note that the length-weighted MDF ρ_l is often referred grain boundary character distribution (GBCD); see, for instance, [13].

From (3), (4) we obtain

$$\frac{\rho_l(\beta,t)}{\rho_n(\beta,t)} = \frac{L(\beta,t)}{\overline{L}(t)} \quad \forall \beta \in [0,\beta_{\max}] \quad \forall t \in [0,T].$$
(5)

The right-hand side of this equality represents a dimensionless average boundary length for a particular misorientation $\beta \in [0,\beta_{\text{max}}]$.

We describe topological transitions accompanying twodimensional grain growth as either neighbor switchings or grain disappearances; see, for instance, [8,16,18]. Assume that initial microstructures do not contain two-sided grains and that only three-sided grains are allowed to disappear. Then, in particular, transformations of three-sided grains into two-sided grains by neighbor switchings are forbidden, and, therefore, two-sided grains cannot appear. This assumption is natural for vertex models [16–18]. Furthermore, the results of Monte Carlo and phase field simulations introduced in Refs. [5,6] show that number fractions of two-sided grains are likely to be small even for sufficiently wide classes of highly anisotropic systems.

Moreover, for the sake of convenience, let us impose an idealizing and simplifying assumption. Suppose that the number of grain boundaries changes with time not discretely but continuously. In this case, $N(\beta,t)$ and N(t) are considered to be real-valued functions. Apply the same assumption to the functions $N_s(t)$ and $N_d(t)$ representing total numbers of neighbor switchings and grain disappearances occurring up to a varying time moment *t*. It is obvious that $N_s(0) = N_d(0) = 0$.

Since in our considerations, the misorientation arguments are continuous rather than discrete, then, more precisely, for every $t \in [0,T]$, $N(\cdot,t)$ and $L(\cdot,t)$ should be interpreted as the measure densities such that the number and the total length of boundaries with misorientations from a Borel set $B \subseteq [0, \beta_{\text{max}}]$ at time *t* are the integrals $\int_{D}^{\infty} N(\beta, t) d\beta$ and $\int_{D}^{\infty} L(\beta, t) d\beta$, respectively. In particular,

$$\int_{0}^{\beta_{\max}} N(\beta, t) d\beta = N(t), \quad \int_{0}^{\beta_{\max}} L(\beta, t) d\beta = L(t) \quad \forall t \in [0, T].$$
(6)

Hence, by taking (4) into account, we conclude that $\rho_n(\cdot,t)$, $\rho_l(\cdot,t)$ are indeed probability densities for all $t \in [0,T]$:

$$\int_{0}^{\beta_{\max}} \rho_n(\beta, t) d\beta = \int_{0}^{\beta_{\max}} \rho_l(\beta, t) d\beta = 1 \quad \forall t \in [0, T].$$
(7)

Our goal is to investigate the number- and length-weighted MDFs (4). Due to (5), the problem of predicting both of them can be reduced to the problem of predicting, for instance, ρ_n and the right-hand side of (5); see also [8]. In what follows, we develop a novel kinetic model for obtaining the number-weighted MDF and estimate the quotient (5) by means of additional arguments.

3. Kinetic equation for the number-weighted MDF

We consider four types of possible critical events:

- creation of a boundary with a particular misorientation as a result of a neighbor switching;
- removal of a boundary with a particular misorientation as a result of a neighbor switching;
- removal of a zero-length boundary with a particular misorientation as a result of a grain disappearance;
- removal of a nonzero-length boundary with a particular misorientation as a result of a grain disappearance.

Adopt the following notations:

- $[0,\beta_{\max}] \Rightarrow \beta \rightarrow P^{s,+}(\beta,t)$ is the probability density for misorientations of boundaries created as a result of neighbor switchings at the moment *t*;
- $[0,\beta_{\max}] \Rightarrow \beta \rightarrow P^{s,-}(\beta,t)$ is the probability density for misorientations of boundaries removed with zero length as a result of neighbor switchings at the moment *t*;
- $[0, \beta_{\text{max}}] \ni \beta \rightarrow P_1^d(\beta, t)$ is the probability density for misorientations of boundaries removed with zero length as a result of grain disappearances at the moment *t*;
- $[0, \beta_{\text{max}}] \ni \beta \rightarrow P_2^d(\beta, t)$ is the probability density for misorientations of boundaries removed with nonzero length as a result of grain disappearances at the moment *t*.

In the next section, the introduced probability densities will be expressed in terms of the sought-for number-weighted MDF.

Also let $\lambda(t)$ be the ratio of the neighbor switchings' rate to the grain disappearances' rate at the moment *t*.

Now let us derive a kinetic equation for the number-weighted MDF. This equation is going to be a modification of the equation introduced in Ref. [8, subsection 2.2]. Fix $\beta \in [0,\beta_{\text{max}}]$, $t \in [0,T]$, and take a real number $\Delta t \in [-t,T-t] \setminus \{0\}$ with a small absolute value. We have

$$N(\beta, t + \Delta t) - N(\beta, t) = \left(P^{s,+}(\beta, t) - P^{s,-}(\beta, t)\right) \cdot \left(N_s(t + \Delta t) - N_s(t)\right) - P_1^d(\beta, t) \cdot \left(N_d(t + \Delta t) - N_d(t)\right) - P_2^d(\beta, t) \cdot 2\left(N_d(t + \Delta t) - N_d(t)\right).$$
(8)

Here we assume that when a three-sided grain disappears, only one of its sides is eliminated with zero length. Due to (4), the left-hand side of (8) can be represented as

$$\begin{split} N(\beta, t + \Delta t) &- N(\beta, t) \approx \frac{\partial N(\beta, t)}{\partial t} \cdot \Delta t \\ &= \frac{\partial}{\partial t} (N(t) \cdot \rho_n(\beta, t)) \cdot \Delta t \\ &= \left(\frac{\partial N(t)}{\partial t} \cdot \rho_n(\beta, t) + N(t) \cdot \frac{\partial \rho_n(\beta, t)}{\partial t} \right) \cdot \Delta t. \end{split}$$

Note that $\frac{\partial N(t)}{\partial t}$, $\frac{\partial \rho_n(\beta,t)}{\partial t}$ are considered to be right partial derivatives for t = 0 and as left partial derivatives for t = T. Divide both sides of (8) by the difference

$$N(t+\Delta t) - N(t) \approx \frac{\partial N(t)}{\partial t} \cdot \Delta t.$$

Since only three-sided grains can disappear at once, then

$$\frac{N_d(t+\Delta t)-N_d(t)}{N(t+\Delta t)-N(t)}=-\frac{1}{3},$$

which implies

$$\begin{split} &\frac{N_s(t+\Delta t)-N_s(t)}{N(t+\Delta t)-N(t)} = \frac{N_s(t+\Delta t)-N_s(t)}{N_d(t+\Delta t)-N_d(t)} \\ & \cdot \frac{N_d(t+\Delta t)-N_d(t)}{N(t+\Delta t)-N(t)} = -\frac{N_s(t+\Delta t)-N_s(t)}{3\cdot(N_d(t+\Delta t)-N_d(t))} \\ & \approx -\frac{\frac{\partial N_s(t)}{\partial t}}{3\cdot\frac{\partial N_d(t)}{\partial t}} = -\frac{\lambda(t)}{3}. \end{split}$$

Finally, we get

ſ

$$p_{n}(\beta,t) + N(t) \cdot \frac{\partial \rho_{n}(\beta,t)}{\partial N(t)}$$

$$= -\frac{\lambda(t)}{3} \cdot \left(P^{s,+}(\beta,t) - P^{s,-}(\beta,t) \right)$$

$$+ \frac{1}{3} \cdot P_{1}^{d}(\beta,t) + \frac{2}{3} \cdot P_{2}^{d}(\beta,t).$$
(9)

Since Eq. (9) does not contain any partial derivatives with respect to *t*, while λ can be considered as a function of *N*, i. e., $\lambda = \lambda(N)$, and $P^{s,+}$, $P^{s,-}$, P_1^d , P_2^d are supposed to be expressed in terms of ρ_n , then here it is possible to replace the dependence on *t* with the dependence on *N*. This is reasonable also because of the fact that N = N(t)is considered to be a strictly decreasing real-valued function; this implies the existence of the inverse function t = t(N). For instance,

$$\frac{dN(t)}{dt} = -g(N(t), t) \quad \forall t \in [0, T],
g(0, t) \equiv 0 \text{ and } g(\eta, t) > 0 \quad \forall \eta > 0 \quad \forall t \in [0, T].$$
(10)

Thus, we come to the equation

$$\rho_{n}(\beta, N) + N \cdot \frac{\partial \rho_{n}(\beta, N)}{\partial N}$$

$$= -\frac{\lambda(N)}{3} \cdot \left(P^{s,+}(\beta, N) - P^{s,-}(\beta, N) \right)$$

$$+ \frac{1}{3} \cdot P_{1}^{d}(\beta, N) + \frac{2}{3} \cdot P_{2}^{d}(\beta, N).$$
(11)

Here

$$N_0 \stackrel{\text{def}}{=} N \Big|_{t=0}, \quad N_{\min} \stackrel{\text{def}}{=} N \Big|_{t=T}$$
 (12)

are parameters. The initial condition is taken as

$$\rho_n(\beta, N_0) = \rho_n^0(\beta) \quad \forall \beta \in [0, \beta_{\max}],$$
(13)

where ρ_n^0 is a specified initial number-weighted MDF.

If (10) holds, then this Cauchy problem has the following form in the variables β ,*t*:

$$\begin{cases} \rho_{n}(\beta,t) - \frac{N(t)}{g(N(t),t)} \cdot \frac{\partial \rho_{n}(\beta,t)}{\partial t} = -\frac{\lambda(t)}{3} \cdot \left(P^{s,+}(\beta,t) - P^{s,-}(\beta,t)\right) \\ + \frac{1}{3} \cdot P_{1}^{d}(\beta,t) + \frac{2}{3} \cdot P_{2}^{d}(\beta,t) \quad \forall \beta \in [0,\beta_{\max}] \quad \forall t \in [0,T], \\ \rho_{n}(\beta,t)|_{t=0} = \rho_{n}^{0}(\beta) \quad \forall \beta \in [0,\beta_{\max}]. \end{cases}$$

$$(14)$$

Let us write a simple backward difference numerical scheme for Cauchy problem (11), (13):

dimensional case.

First, introduce the auxiliary function

. .

$$[-2\beta_{\max}, 2\beta_{\max}] \ni \xi \to \omega(\xi) \stackrel{\text{def}}{=} \begin{cases} |\xi|, & |\xi| \in [0, \beta_{\max}], \\ 2\beta_{\max} - |\xi|, & |\xi| \in [\beta_{\max}, 2\beta_{\max}]. \end{cases}$$
(16)

It transforms angles from $[-2\beta_{max}, 2\beta_{max}]$ to angles from $[0, \beta_{max}]$ according to the symmetry with respect to zero and $(2\beta_{max})$ -periodicity of misorientations, which together imply, in particular, the symmetry with respect to β_{max} on $[0, 2\beta_{max}]$.

According to Figs. 1–4, for all $\beta \in [0,\beta_{max}]$ and $N \in [N_{\min},N_0]$, $P^{s,+}(\beta,N)$, $P^{s,-}(\beta,N)$, $P^d_1(\beta,N)$, $P^d_2(\beta,N)$ can be represented in terms of the number-weighted MDF as

$$P^{s,+}(\beta, N) = \frac{p^{s,+}(\beta, N)}{\int_{0}^{\beta_{\max}} p^{s,+}(\xi, N) d\xi},$$

$$P^{s,-}(\beta, N) = \frac{p^{s,-}(\beta, N)}{\int_{0}^{\beta_{\max}} p^{s,-}(\xi, N) d\xi},$$

$$P_{1}^{d}(\beta, N) = \frac{p_{1}^{d}(\beta, N)}{\int_{0}^{\beta_{\max}} p_{1}^{d}(\xi, N) d\xi},$$

$$P_{2}^{d}(\beta, N) = \frac{p_{2}^{d}(\beta, N)}{\int_{0}^{\beta_{\max}} p_{2}^{d}(\xi, N) d\xi},$$
(17)

$$\begin{cases} \rho_n(\beta, N - \Delta N) = \rho_n(\beta, N) + \frac{\Delta N}{N} \cdot \left(\rho_n(\beta, N) + \frac{\lambda(N)}{3} \cdot \left(P^{s,+}(\beta, N) - P^{s,-}(\beta, N) \right) - \frac{1}{3} \cdot P_1^d(\beta, N) - \frac{2}{3} \cdot P_2^d(\beta, N) \right), \quad \Delta N > 0 \text{ is sufficiently small,} \\ \rho_n(\beta, N_0) = \rho_n^0(\beta). \end{cases}$$

$$(15)$$

4. Representations of $P^{s,+}$, $P^{s,-}$, P^d_1 , P^d_2

We propose a modification and extension of the onedimensional approach introduced in Ref. [24] to our two-



Fig. 1. To formula (18). We focus on configurations before critical events.



Fig. 2. To formula (19). We focus on configurations before critical events.

where



Fig. 3. To formula (20). We focus on configurations before critical events.

		4
$p^{\mathbf{S}+}(\beta, \mathbf{N}) \stackrel{\text{def}}{=} \iint_{(\beta_1, \beta_2) \in [-\beta_{\max}, \beta_{\max}] \times [-\beta_{\max}, \beta_{\max}]: \gamma(\omega(\beta_2 - \beta_1)) \geqslant \gamma(\beta)} \rho_{\mathbf{n}}(\omega(\beta_2 - \beta_1), \mathbf{N}) \cdot \rho_{\mathbf{n}}(\beta)}$	$(\beta_1 ,N)\cdot\rho_n(\beta_2 ,N)\cdot\rho_n(\omega(\beta+\beta_1),N)\cdot\rho_n(\omega(\beta+\beta_2),N)d\beta_1d\beta_2,$	(18)
$p^{S^{-}}(\beta, N) \stackrel{def}{=} \rho_{n}(\beta, N) \cdot \underbrace{\iint}_{(\beta_{1}, \beta_{2}) \in [-\beta_{\max}, \beta_{\max}] \times [-\beta_{\max}, \beta_{\max}] : \gamma(\omega(\beta_{2} - \beta_{1})) \leq \gamma(\beta)} \rho_{n}(\beta_{1} , N) \cdot \rho_{n}(\beta_{1}) + \beta_{\max}(\beta_{1} - \beta_{\max}) \cdot \beta_{\max}(\beta_{1} - \beta_{\max}) + \beta_{\max}(\beta_{1} -$	$\rho_{n}(\beta_{2} ,N)\cdot\rho_{n}(\omega(\beta+\beta_{1}),N)\cdot\rho_{n}(\omega(\beta+\beta_{2}),N)d\beta_{1}d\beta_{2},$	(19)
$p_1^{d}(\beta, \mathbf{N}) \stackrel{def}{=} \rho_{n}(\beta, \mathbf{N}) \cdot \iint_{(\beta_1, \beta_2) \in [-\beta_{\max, \beta_{\max}, \beta_{\max}, \beta_{\max}, \beta_{\max}] \times [-\beta_{\max}, \beta_{\max}] \times [\beta_0]} \iint_{\gamma(\beta_0), \gamma(\beta_1) + \gamma(\beta_2)}$	$\rho_n(\beta_1 ,N)\cdot\rho_n(\beta_2 ,N)\cdot\rho_n(\omega(\beta+\beta_1),N)\cdot\rho_n(\omega(\beta+\beta_2),N)\cdot\rho_n(\omega(\beta_2-\beta_1),N)d\beta_1d\beta_2,\\ \beta_2)\}$	(20)

(21)

 $\rho_{n}(|\beta_{1}|,N)\cdot\rho_{n}(|\beta_{2}|,N)\cdot\rho_{n}(\omega(\beta+\beta_{1}),N)\cdot\rho_{n}(\omega(\beta+\beta_{2}),N)\cdot\rho_{n}(\omega(\beta_{2}-\beta_{1}),N)d\beta_{1}d\beta_{2}$

 $(\beta_1,\beta_2) \in [-\beta_{\max},\beta_{\max}] \times [-\beta_{\max},\beta_{\max}]: \widetilde{\gamma(\omega(\beta+\beta_2))} \leqslant \min \{\gamma(|\beta_1|),\gamma(|\beta_1|)+\gamma(|\beta_2|)\}$

 $p_2^d(\beta, N) \stackrel{\text{def}}{=} \rho_n(\beta, N)$.

 $(\beta_1,\beta_2) \in [-\beta_{\max},\beta_{\max}] \times [-\beta_{\max},\beta_{\max}]; \gamma(\omega(\beta_2-\beta)) \leqslant \min \left\{ \gamma(|\beta_1|),\gamma(|\beta|) + \gamma(|\beta_2|) \right\}$

+

 $\rho_{n}(|\beta_{1}|,N) \cdot \rho_{n}(|\beta_{2}|,N) \cdot \rho_{n}(\omega(\beta_{1}-\beta),N) \cdot \rho_{n}(\omega(\beta_{2}-\beta),N) \cdot \rho_{n}(\omega(\beta_{2}-\beta_{1}),N) \ d\beta_{1}d\beta_{2}$



Fig. 4. To formula (21). We focus on configurations before critical events.

In (18)–(21), we determine the sets of integration by inequalities specifying the energy dissipation as a result of topological transitions. Note that we consider every grain disappearance as a combination of the following two consecutive events which should be energy dissipative:

- the neighbor switching which removes one zero-length boundary of a three-sided grain, generates a new zero-length boundary, and binds the remaining two boundaries of the destroyed grain into a two-sided grain;
- the instantaneous removal of the generated zero-length boundary together with the whole constructed two-sided grain.

Such interpretation is stipulated by the convention that after a grain disappearance, a new triple junction is located at the first collision point [18]. Besides, as in Ref. [24], we suppose the absence of correlations when writing the right-hand sides of (18)-(21). In order to be convinced that such assumption is reasonable and not too restrictive, we have to compare numerical results of our model with those obtained for the same initial data by some generally recognized large-scale simulation technique, for instance, by the phase field model described in Refs. [9–11]. This will be done in Section 7.

5. Important remarks

Remark 1. Consider the case when the initial number-weighted grain orientation distribution is uniform on the interval $[0,\alpha_{\max}] = [0,2\beta_{\max}]$ (recall (1)). Let $\tilde{\rho}_n^0$ denote the probability density for the absolute value of the difference of two independent random variables equidistributed on $[0,\alpha_{\max}]$. Then

$$\tilde{\rho}_{n}^{0}(\beta) = \begin{cases} \frac{2(\alpha_{\max} - \beta)}{\alpha_{\max}^{2}}, & \beta \in [0, \alpha_{\max}], \\ 0, & \beta \in \mathbb{R} \setminus [0, \alpha_{\max}]. \end{cases}$$

By taking the misorientations' symmetry with respect to β_{max} on $[0, \alpha_{max}]$ into account, we conclude that the initial number-

weighted MDF
$$\rho_n^{0 \stackrel{\text{def}}{=}} \rho_n \Big|_{N=N_0}$$
 is uniform on [0, β_{max}]:

$$\rho_n^0(\beta) = \tilde{\rho}_n^0(\beta) + \tilde{\rho}_n^0(\alpha_{\max} - \beta) = \frac{1}{\beta_{\max}} \quad \forall \beta \in [0, \beta_{\max}],$$

$$\rho_n^0(\beta) = \begin{cases} \frac{1}{\beta_{\max}}, & \beta \in [0, \beta_{\max}], \\ 0, & \beta \in \mathbb{R} \setminus [0, \beta_{\max}]. \end{cases}$$
(22)

In addition to (22), it is natural to assume that the initial length-

weighted MDF $\rho_l^{\text{odef}} = \rho_l \Big|_{N=N_0}$ is uniform on [0, β_{max}], i. e.,

$$\rho_l^0(\beta) = \begin{cases} \frac{1}{\beta_{\max}}, & \beta \in [0, \beta_{\max}], \\ 0, & \beta \in \mathbb{R} \setminus [0, \beta_{\max}]. \end{cases}$$
(23)

Indeed, for the large-scale simulations within the framework of the papers [10,20], the generated initial fiber textures fulfill (22), (23) with a fairly good degree of accuracy.

Remark 2. By the use of (18), (19), one can easily verify the equality

$$\frac{\rho_n(\beta, N - \Delta N) - \rho_n(\beta, N)}{\Delta N} = A(\beta, N) \cdot \rho_n(\beta, N) + B(\beta, N),$$
(25)

where

$$A(\beta, N) \stackrel{\text{def}}{=} \frac{1}{N} \cdot \left(1 - \frac{\lambda(N)}{3} \cdot \frac{P^{s, -}(\beta, N)}{\rho_n(\beta, N)} - \frac{1}{3} \cdot \frac{P_1^d(\beta, N)}{\rho_n(\beta, N)} - \frac{2}{3} \cdot \frac{P_2^d(\beta, N)}{\rho_n(\beta, N)} \right),$$
$$B(\beta, N) \stackrel{\text{def}}{=} \frac{1}{N} \cdot \frac{\lambda(N)}{3} \cdot P^{s, +}(\beta, N).$$
(26)

In (26), the quotients $\frac{p^{s.-}(\beta,N)}{\rho_n(\beta,N)}$, $\frac{P_1^d(\beta,N)}{\rho_n(\beta,N)}$, $\frac{P_2^d(\beta,N)}{\rho_n(\beta,N)}$ are understood as the corresponding coefficients near $\rho_n(\beta,N)$ in (19)–(21); such interpretation is valid, in particular, for $\rho_n(\beta,N) = 0$. Consider (25) as an approximation of a linear one-dimensional ordinary differential equation with constant coefficients on the interval $[0,\Delta N]$ for a sufficiently small $\Delta N > 0$:

$$\frac{\partial \rho_n(\beta, N + (-1) \cdot h)}{\partial h} = A(\beta, N) \cdot \rho_n(\beta, N + (-1) \cdot h) + B(\beta, N), \quad h \in [0, \Delta N].$$
(27)

Then we can propose a numerical scheme based on a formula for the exact solution of (27):

$$\rho_{n}(\beta, N - \Delta N) = \begin{cases} \rho_{n}(\beta, N) \cdot e^{A(\beta, N) \cdot \Delta N} + \frac{B(\beta, N)}{A(\beta, N)} \cdot \left(e^{A(\beta, N) \cdot \Delta N} - 1 \right), & A(\beta, N) \neq 0, \\ \rho_{n}(\beta, N) + B(\beta, N) \cdot \Delta N, & A(\beta, N) = 0, \\ = \rho_{n}(\beta, N) \cdot e^{A(\beta, N) \cdot \Delta N} + B(\beta, N) \cdot \left(\Delta N + \sum_{k=2}^{\infty} \frac{(A(\beta, N))^{k-1} \cdot (\Delta N)^{k}}{k!} \right), & \Delta N > 0 \text{ is sufficiently small}, \rho_{n}(\beta, N_{0}) = \rho_{n}^{0}(\beta). \end{cases}$$

$$(28)$$

$$\int_{0}^{\beta_{\max}} p^{s,+}(\beta,N) d\beta = \int_{0}^{\beta_{\max}} p^{s,-}(\beta,N) d\beta \quad \forall N \in [N_{\min},N_0],$$
(24)

which conforms with the fact that each neighbor switching removes a single boundary and creates one new boundary.

Remark 3. Now consider the first equality in numerical scheme (15) and suppose that the denominators in the right-hand sides of (17) do not vanish. If at some point $(\beta,N) \in [0,\beta_{max}] \times [N_{min} + \Delta N, N_0]$, $\rho_n(\beta, N) = 0$ and also $\rho_n(\xi, N) \ge 0$ for all $\xi \in [0,\beta_{max}]$, then due to (17)-(21), $P^{s,+}(\beta,N) \ge 0$ and $P^{s,-}(\beta,N) = P_1^d(\beta,N) = P_2^d(\beta,N) = 0$, which leads to $\rho_n(\beta,N - \Delta N) \ge 0$. Therefore, the solution to Cauchy problem (11), (13) should not take negative values if $\rho_n^0(\xi) \ge 0$ for all $\xi \in [0,\beta_{max}]$ and $\int_0^{\beta_{max}} \rho_n^0(\xi) d\xi > 0$. Furthermore, by taking (15), (17) into account, one can easily verify that if $\int_0^{\beta_{max}} \rho_n(\xi,N) d\xi = 1$

then
$$\int_0^{\rho_{\max}} \rho_n(\xi, N - \Delta N) d\xi = 1.$$

Remark 4. In order to exclude negative values of an approximate solution to Cauchy problem (11), (13) which are possible in case of sufficiently large computational inaccuracies, we need to transform numerical scheme (15). The first equality in (15) can be written as

For this scheme, the a priori order of accuracy is not less than for (15). Besides, an approximate solution obtained by (28) should not take negative values for nonnegative initial data, because in this case, $B(\beta,N) \ge 0$ due to (17), (18), (26), and

$$\frac{1}{A(\beta,N)} \cdot \left(e^{A(\beta,N) \cdot \Delta N} - 1 \right) > 0 \quad \text{for } A(\beta,N) \in \mathbb{R} \setminus \{0\}.$$

Remark 5. Due to (17)–(21), the probability densities $P^{s,+}$, $P^{s,-}$, P_1^d , P_2^d explicitly depend not on the grain boundary energy density (2) itself but on the corresponding relation of order. However, the ratio λ of the neighbor switchings' rate to the grain disappearances' rate is expected to depend directly on γ .

Note that in Ref. [34] λ was obtained from large-scale simulations via a two-dimensional curvature-driven vertex model applied to an initially Monte Carlo generated microstructure. Moreover, a concise and informal discussion of possible ranges for λ was given in Ref. [35]. Nevertheless, by the time of working on this paper, we had not seen any general approaches to estimate λ described in the literature on mathematical modeling of grain growth. Let us propose our own heuristic method to estimate λ .

For every $N \in [N_{\min}, N_0]$, introduce the average grain boundary energy density and the fractions of low-energy and high-energy boundaries:

$$\overline{\gamma}(N) \stackrel{\text{def}}{=} \int_{0}^{\beta_{\text{max}}} \gamma(\beta) \cdot \rho_{n}(\beta, N) \ d\beta,$$

$$\kappa_{\text{le}}(N) \stackrel{\text{def}}{=} \int_{\beta \in [0, \beta_{\text{max}}]: \gamma(\beta) \leq \overline{\gamma}(N)} \rho_{n}(\beta, N) \ d\beta, \quad \forall N \in [N_{\min}, N_{0}]. \quad (29)$$

$$\kappa_{\text{he}}(N) \stackrel{\text{def}}{=} \int_{\beta \in [0, \beta_{\text{max}}]: \gamma(\beta) \geq \overline{\gamma}(N)} \rho_{n}(\beta, N) \ d\beta$$

Here grain boundaries with energy densities less or equal to the average value $\overline{\gamma}(N)$ are called low-energy, whereas grain boundaries with energy densities greater or equal to $\overline{\gamma}(N)$ are called high-energy. When $N = N_0$, i. e., t = 0, according to (22), we have

$$\begin{aligned} \overline{\gamma}_{0} \stackrel{\text{def}}{=} \overline{\gamma}(N_{0}) &= \frac{1}{\beta_{\max}} \cdot \int_{0}^{\beta_{\max}} \gamma(\beta) d\beta, \\ \kappa_{\text{le}}(N_{0}) &= \frac{1}{\beta_{\max}} \cdot \mu(\{\beta \in [0, \beta_{\max}] : \gamma(\beta) \leq \overline{\gamma}_{0}\}), \\ \kappa_{\text{he}}(N_{0}) &= \frac{1}{\beta_{\max}} \cdot \mu(\{\beta \in [0, \beta_{\max}] : \gamma(\beta) \geq \overline{\gamma}_{0}\}), \end{aligned}$$
(30)

where μ denotes the Lebesgue measure on \mathbb{R} .

In the isotropic case when $\gamma(\beta) \equiv \overline{\gamma}_0$ for all $\beta \in [0,\beta_{max}]$, we obtain that $\overline{\gamma}(N) \equiv \overline{\gamma}_0$ and $\kappa_{le}(N) \equiv \kappa_{he}(N) \equiv 1$ for every $N \in [N_{\min},N_0]$. Suppose that in this isotropic case, the ratio $\lambda = \lambda(N)$ of the neighbor switchings' rate to the grain disappearances' rate is equal to some constant $\lambda_{isotr} > 0$ for all $N \in [N_{\min},N_0]$. For instance, in the computer simulations of A. P. Sprague, B. R. Patterson, and S. Grandhi [34] conducted for the isotropic case, the ratio of the cumulative number of neighbor switchings to the cumulative number of grain disappearances turned out to be nearly constant at 1.34, which implies that $\lambda_{isotr} \approx 1.34$ for these simulations.

For anisotropic grain boundary energy densities, we estimate $\boldsymbol{\lambda}$ as

$$\lambda(N) \approx \lambda_{\text{isotr}} \cdot \nu \left(\frac{2 \cdot \kappa_{\text{he}}(N)}{\kappa_{\text{le}}(N) + \kappa_{\text{he}}(N)} \right) \quad \forall N \in [N_{\text{min}}, N_0],$$
(31)

where $\nu : (0,2) \rightarrow (0,+\infty)$ is an increasing function such that $\nu(1) = 1$; for instance, $\nu(x) = x$ for all $x \in (0,2)$. Here we adopt the assumption that $\lambda(N)$ and the quotient

$$\frac{2 \cdot \kappa_{he}(N)}{\kappa_{le}(N) + \kappa_{he}(N)} = \frac{\kappa_{he}(N)}{\frac{1}{2}(\kappa_{le}(N) + \kappa_{he}(N))}$$

should increase or decrease together. This is based on the following natural heuristic hypotheses:

• a grain boundary removed as a result of a neighbor switching is more likely to be high-energy;

• a grain boundary removed with zero length as a result of a grain disappearance is more likely to be high-energy, while the boundary issuing from the corresponding opposite vertex of the shrinking three-sided grain and not belonging to the latter is more likely to be low-energy.

The condition $\nu(1) = 1$ is stipulated by the fact that in the mentioned isotropic case, we have $\kappa_{\text{le}}(N) \equiv \kappa_{\text{he}}(N) \equiv \frac{2 \cdot \kappa_{\text{he}}(N)}{\kappa_{\text{le}}(N) + \kappa_{\text{he}}(N)} \equiv 1$ for every $N \in [N_{\min}, N_0]$.

Finally, note that for a fixed function γ and for functions λ taking values from a sufficiently wide realistic range (i. e., for not too small and not too large λ), the qualitative portraits of the number-weighted MDFs are not expected to differ from each other substantially, especially at final stages of texture evolutions, and the same for the length-weighted MDFs; the results of our numerical simulations conform with such expectations (see remark 1 in Section 7). Indeed, the general energy reduction principles stimulating grain growth do not seem to be directly related to the quotient λ , although the rates of textures' tending to the resulting configurations may depend on λ . Besides, λ is unlikely to grow or decrease infinitely during texture evolutions, because the greater/smaller the high-energy boundaries' fraction is, the greater/smaller the probabilities of their eliminations are expected to be (see (17), (19)–(21)).

6. Representation of the length-weighted MDF

In order to estimate the right-hand side of (5) and, consequently, the length-weighted MDF ρ_{l} , let us apply a heuristic quantitative model which describes the mechanism of low-energy boundaries' lengthening at triple junctions [4,8,36] simplistically. The phenomenon of low-energy boundaries' lengthening was justified experimentally [37].

As in Ref. [36], consider a two-dimensional geometrical stencil where one of the ends of a grain boundary with a scalar misorientation $\beta \in [0,\beta_{max}]$ and, therefore, with the energy density $\gamma(\beta)$ meets two other boundaries having the average energy density $\overline{\gamma}(N)$. Note that in Refs. [4,8] Read-Shockley type energy densities depending on the disorientation variable were considered, and for similar boundary lengthening models, $\overline{\gamma}(N)$ was replaced with the maximum possible value of the energy density. However, taking $\overline{\gamma}(N)$ here seems to be more accurate. To begin, the three boundaries are in an isotropic configuration, i. e., they have the same length equal to the average length $\overline{L}(N)$ and constitute three angles equal to $\frac{2\pi}{3} = 120^{\circ}$. Now we let the boundaries relax until the equilibrium triple line geometry is reached and calculate the resulting length of the first boundary having the energy density $\gamma(\beta)$. Then, under the assumption that

$$\overline{\gamma}(N) > 0 \quad \forall N \in [N_{\min}, N_0], \tag{32}$$

we come to the following simplified estimate for the quotient (5):

$$\frac{1}{c(N)} \cdot \frac{\overline{L}(\beta, N)}{\overline{L}(N)} \approx \begin{cases} 1 + \frac{1}{2} \left(1 - \frac{\sqrt{3}}{\tan\left(\arccos\left(\frac{\gamma(\beta)}{2\overline{\gamma}(N)}\right)}\right) \right), & \frac{\gamma(\beta)}{\overline{\gamma}(N)} \leq \sqrt{3}, \\ 0, & \frac{\gamma(\beta)}{\overline{\gamma}(N)} > \sqrt{3}, \end{cases}$$
(33)

The right-hand side of (33) tends to $\frac{3}{2}$ when $\gamma(\beta) \rightarrow +0$, equals 1 when $\gamma(\beta) = \overline{\gamma}(N)$, and vanishes if $\frac{\gamma(\beta)}{\overline{\gamma}(N)} \ge \sqrt{3}$. Here $N_{\text{interm}} \in (N_{\min}, N_0)$ is some constant; for instance, $N_{\text{interm}} = \frac{N_{\min}+N_0}{2}$. Its presence is called forth by the fact that for anisotropic grain boundary energy densities and the uniform initial number- and length-weighted MDFs (22), (23), relation (33) is not expected to hold at initial stages of texture evolutions. Moreover, for every $N \in [N_{\min}, N_{\text{interm}}]$ the number c(N) is determined from the normalization condition:

$$\int_{0}^{\beta_{\max}} \rho_n(\beta, N) \cdot \frac{\overline{L}(\beta, N)}{\overline{L}(N)} d\beta = 1$$

It is also necessary to mention that for highly anisotropic boundary energy densities, higher-order junctions such as quadruple junctions may become stable [4,6,11,18,20]. Our model does not explicitly operate with stable junctions of orders greater than 3, although we can interpret a quadruple junction as two triple junctions connected through a common grain boundary with a relatively small length. Nevertheless, taking such short boundaries into account or neglecting them may substantially affect the number-weighted MDF but not the length-weighted MDF.

7. Numerical simulations

Recall that we assume the uniformity of the initial number- and length-weighted MDFs according to (22), (23).

First, consider the isotropic case when $\gamma(\beta) \equiv \text{const}$ for every $\beta \in [0,\beta_{\text{max}}]$. Then with the help of (11), (13), (17)–(22), one can easily verify that $\frac{\partial p_n(\beta,N)}{\partial N} = 0$ for all $\beta \in [0,\beta_{\text{max}}]$, $N \in [N_{\min},N_0]$, and that, hence, the number-weighted MDF remains uniform during the whole observed process. Furthermore, in this case, the right-hand side of (33) equals 1 for all $\beta \in [0,\beta_{\text{max}}]$, $N \in [N_{\min},N_0]$, and it is reasonable to choose $N_{\text{interm}} = N_0$, which leads to the uniformity of the length-weighted MDF for every $N \in [N_{\text{min}},N_0]$.

Now consider three nontrivial types of boundary energy densities (see Fig. 5).

Type I. Read-Shockley boundary energy density [2,4,7,10,20,22]:

Type II. Modification of γ_I with an additional local minimum [10]:

Type III. Concave parabolic boundary energy density [16]:

$$[0, \beta_{\max}] \ni \beta \to \gamma_{III}(\beta) \stackrel{\text{def}}{=} \gamma_{l,2} + (\gamma_m) - \gamma_{l,2} \cdot \frac{\beta \cdot (2\beta_{\max} - \beta)}{\beta_{\max}^2}, \quad \gamma_{l,2} \in (0, \gamma_m).$$
(36)

As in Ref. [10], we take

$$\alpha_{\max} = \frac{\pi}{2} = 90^{\circ}, \quad \beta_{\max} = \frac{\pi}{4} = 45^{\circ},$$

$$\gamma_l = 0, \quad \gamma_m = 0.25, \quad \gamma_{l,1} = 0.1,$$

(37)

$$\beta_{\rm I} = \frac{\pi}{12} = 15^{\circ}, \quad \beta_{\rm II} = \frac{5\pi}{24} = 37.5^{\circ}, \quad \beta_{\rm II}' = \frac{\pi}{18} = 10^{\circ}.$$

Besides, for γ_{III} , we choose

$$\gamma_{l,2} = \frac{\gamma_m}{2} = 0.125. \tag{38}$$

For each of the introduced boundary energy densities $\gamma_{\rm I}$, $\gamma_{\rm II}$, $\gamma_{\rm III}$, we consider the following two ways of determining the ratio $\lambda = \lambda(N)$ of the neighbor switchings' rate to the grain disappearances' rate (see remark 5 in Section 5 and recall that the value 1.34 was obtained in Ref. [34]):

A) $\lambda(N) \equiv 1.34$; B) λ is estimated according to (31) with $\lambda_{isotr} = 1.34$ and $\nu(x) \equiv x$.

We take $N_0 = 300,000$; this is reasonable if the initial number of grains is 100,000 and the initial average number of sides per

$$[0,\beta_{\max}] \ni \beta \to \gamma_{l}(\beta) \stackrel{\text{def}}{=} \begin{cases} \gamma_{l}, & \beta = 0, \\ \gamma_{l} + (\gamma_{m} - \gamma_{l}) \cdot \frac{\beta}{\beta_{l}} \cdot \left(1 - \ln\left(\frac{\beta}{\beta_{l}}\right)\right), & \beta \in [0,\beta_{l}], & \gamma_{l} \ge 0, & \gamma_{m} > \gamma_{l}, & \beta_{l} \in [0,\beta_{\max}]. \\ \gamma_{m}, & \beta \ge \beta_{l}, \end{cases}$$
(34)

 $[N_{\min},N_0]$. Thus, our model gives natural and trivial results under initial conditions (22), (23) if the boundary energy density is constant.

grain is 6, because each grain boundary belongs to exactly two grains (see, for instance, [33, section 4.6.2]). In order to compute ρ_n , we apply numerical scheme (28) with $\Delta N = 7500$ step-by-step till

$$[\mathbf{0},\beta_{\max}] \ni \beta \to \gamma_{\mathrm{II}}(\beta) \stackrel{\text{def}}{=} \begin{cases} \gamma_{\mathrm{I}}(\beta), \quad \beta \in [\mathbf{0},\beta_{\mathrm{II}} - \beta'_{\mathrm{II}}], \\ \gamma_{l,1} + (\gamma_{m} - \gamma_{l,1}) \cdot \frac{|\beta - \beta_{\mathrm{II}}|}{\beta'_{\mathrm{II}}} \cdot \left(1 - \ln\left(\frac{|\beta - \beta_{\mathrm{II}}|}{\beta'_{\mathrm{II}}}\right)\right), \\ \beta \in [\beta_{\mathrm{II}} - \beta'_{\mathrm{II}}, \quad \beta_{\mathrm{II}} + \beta'_{\mathrm{II}}] \setminus \{\beta_{\mathrm{II}}\}, \\ \gamma_{l,1}, \quad \beta = \beta_{\mathrm{II}}, \\ \gamma_{m}, \quad \beta \geqslant \beta_{\mathrm{II}} + \beta'_{\mathrm{II}}, \end{cases} \qquad \gamma_{l,1} \in (\gamma_{l}, \gamma_{m}), \quad \beta_{\mathrm{II}} \in (\beta_{l}, \beta_{\max}], \quad \beta'_{\mathrm{II}} \in (0, \beta_{\mathrm{II}} - \beta_{\mathrm{II}}]. \tag{35}$$



the very end at which $\rho_n|_{N=0}$ is obtained by the use of $\rho_n|_{N=\Delta N}$. When having an approximate ρ_n , we apply (33) with $N_{\text{interm}} = \frac{N_0}{2} = 150\,000$ so as to estimate ρ_l from $N = N_{\text{interm}}$ to N = 0.

The results of our numerical simulations are presented in Figs. 6–20. The following remarks can be made.

Remark 1. From Figs. 6–9, 14–19, we see only a slight difference

between cases **A** and **B**, which conforms with our hypothesis at the end of remark 5 in Section 5.

Fig. 6. Number-weighted MDF in case A.

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Remark 2. Figs. 6–9 show that:









Fig. 7. Number-weighted MDF in case B.



Length-weighted MDF for the boundary energy density of Type II and $\lambda(N) \equiv 1.34$



Length-weighted MDF for the boundary energy density of Type III and $\lambda(N) \equiv 1.34$



Fig. 8. Length-weighted MDF in case A.



Fig. 9. Length-weighted MDF in case B.

Fig. 10. Evolution of probability density $P^{s,+}(\cdot,N)$ in case **B**.

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Fig. 11. Evolution of probability density $P^{s,-}(\cdot,N)$ in case **B**.



Evolution of probability density P_1^d for the boundary energy density of Type II and $\lambda(N)$ determined by (30) with $\lambda_{isotr} \equiv 1.34$ and $\nu(x) \equiv x$



Evolution of probability density P_1^d for the boundary energy density of Type III and $\lambda(N)$ determined by (30) with $\lambda_{isotr} \equiv 1.34$ and $\nu(x) \equiv x$







Evolution of probability density P_2^d for the boundary energy density of Type II and $\lambda(N)$ determined by (30) with $\lambda_{isotr} \equiv 1.34$ and $v(x) \equiv x$





Fig. 13. Evolution of probability density $P_2^d(\cdot, N)$ in case **B**.



The right-hand side of (33) for the boundary energy density of Type II and $\lambda(N)\equiv 1.34$



The right-hand side of (33) for the boundary energy density of Type III and $\lambda(N) \equiv 1.34$



• for the boundary energy density γ_1 with a plateau stretching from $\beta = \beta_1 = \frac{\pi}{12} = 15^\circ$ to $\beta = \beta_{max} = \frac{\pi}{4} = 45^\circ$, the number-and length-weighted MDFs are discontinuous at $\beta = \beta_1$;













Evolution of low-energy and high-energy boundaries'





Evolution of low-energy and high-energy boundaries' fractions in case of the boundary energy density of $T_{\rm res}$ ILL and $\lambda(N) = 1.24$



Fig. 16. Evolution of low-energy and high-energy boundaries' fractions $\kappa_{le}(N)$, $\kappa_{he}(N)$ in case **A**.

Evolution of low-energy and high-energy boundaries' fractions in case of the boundary energy density of Type I and $\lambda(N)$ determined by (30) with $\lambda_{isotr} = 1.34$ and $\nu(x) \equiv x$







Evolution of low-energy and high-energy boundaries' fractions in case of the boundary energy density of Type III and $\lambda(N)$ determined by (30) with $\lambda_{isotr} = 1.34$ and $\nu(x) \equiv x$



Fig. 17. Evolution of low-energy and high-energy boundaries' fractions $\kappa_{le}(N)$, $\kappa_{he}(N)$ in case **B**.

- for the boundary energy density γ_{II} with a plateau stretching from $\beta = \beta_I$ to $\beta = \beta_{II} - \beta'_{II} = \frac{11\pi}{72} = 27.5^\circ$, the number- and length-weighted MDFs are discontinuous at $\beta = \beta_I$ and $\beta = \beta_{II} - \beta'_{II}$;
- for the boundary energy density γ_{III} without any plateaus, the number- and length-weighted MDFs are continuous.

Indeed, for $\gamma_{\rm I},\gamma_{\rm II}$, the probability densities $P^{\rm S,-}(\cdot,N)$, $P_1^d(\cdot,N)$, $P_2^d(\cdot,N)$ turn out to be discontinuous at the corresponding misorientations, which is not detected for $\gamma_{\rm III}$; see Figs. 10–13 for case **B**, a similar situation is observed in case **A**. This is related to the energy inequalities specifying the sets of integration in (19), (20). For misorientations β near an energy plateau but not on it, these sets of integration are smaller than for plateau misorientations, where $P^{\rm S,-}(\cdot,N)$ and $P_1^d(\cdot,N)$ therefore have wells at the beginning of the evolutions, i. e., for sufficiently large *N*. Subsequently, $P_2^d(\cdot,N)$ also becomes discontinuous, and the directions of the jumps of $P^{\rm S,-}(\cdot,N)$ and $P_1^d(\cdot,N)$ change according to the directions of the jumps of proportional to $\rho_n(\beta,N)$. Moreover, $\rho_n(\cdot,N)$ does not change the directions of its jumps.

These discontinuities are not seen explicitly on the histograms for the number- and length-weighted MDFs demonstrated in [10, Figures 8(a,b), 9(a,b)] and [20, Figure 3(b)] as applied to boundary energy densities of Types I, II (note that in Ref. [20] the Read-Shockley boundary energy density with the other parameters $\gamma_l = 0.1$ and $\gamma_m = 0.9$ was chosen). The possible deficiency of enabling such discontinuities is outweighed by the fact that our approach is rather general and relies on a smaller amount of properties and parameters than considered, for instance, by largescale phase field models described in Refs. [6,9–11]. In addition, this drawback does not seem to be crucial, because the detected discontinuities of the number- and length-weighted MDFs are sufficiently small and should not essentially violate the natural structures of the MDFs.

Besides, jumps of the number-weighted MDF with respect to the misorientation variable should not strongly interfere with using numerical scheme (28), since Eq. (11) contains a partial derivative only with respect to *N*. From Figs. 6–20, some computational inaccuracies can be noticed closer to the end N = 0, but they do not look significant.

Remark 3. From Figs. 6–9, we observe a qualitative similarity with the simulation results presented in Refs. [10, section 3.3, Figures 8(a,b), 9(a,b)] and [20, Figure 3(b)] for boundary energy densities of Types I, II (Type III was not considered in Refs. [10,20]).

Recall that our model is not exactly kinetic, because it operates with the decreasing total number of grain boundaries instead of the time variable. This somewhat complicates the direct comparison. However, important common tendencies are clearly seen. For γ_1 (similarly for γ_{III}), $\rho_n(\cdot,N)$ and $\rho_l(\cdot,N)$ get a single peak at $\beta = 0$ corresponding to the unique local and simultaneously absolute minimum of this boundary energy density. For γ_{II} , an extra peak occurs at $\beta = \beta_{II}$, where γ_{II} has its second local minimum which is not absolute. For sufficiently large *N*, the left peak grows slightly faster than the right one. For small *N*, the left peak grows much faster than the right one.

Although the quantitative differences between our numerical results and what is shown in Refs. [10, section 3.3, Figures 8(a,b), 9(a,b), [20, Figure 3(b)] do not seem to be crucial, we do not obtain that for γ_{II} the final left peaks of the MDFs are substantially smaller than for γ_{I} . Nevertheless, we do not know any physical reasons explaining why there should be such kind of distinction. Besides, in Ref. [10, section 3.3], it was noticed that the heights of the left and right peaks of the MDFs should be quite sensitive to the number of possible discrete misorientations considered by phase field models.







Evolution of the energy density averaged over all existing boundaries in the system (Type III of the energy dependence on misorientations, $\lambda(N) \equiv 1.34$)



Fig. 18. Evolution of the average energy density $\overline{\gamma}(N)$ in case **A**.

In particular, the greater this number, the higher the left peaks are likely to be.

Also note that according to Ref. [10, section 3.1], N. Moelans, F. A.



Evolution of the energy density averaged over all

Evolution of the energy density averaged over all existing boundaries in the system (Type II of the energy dependence on misorientations, $\lambda(N)$ determined by (30) with $\lambda_{isotr} = 1.34$ and $v(x) \equiv x$)



Evolution of the energy density averaged over all existing boundaries in the system (Type III of the energy dependence on misorientations, $\lambda(N)$ determined by (30) with $\lambda_{isotr} = 1.34$ and $\nu(x) \equiv x$)



Fig. 19. Evolution of the average energy density $\overline{\gamma}(N)$ in case **B**.











Fig. 20. Evolution of the ratio $\lambda(N)$ of the neighbor switchings' rate to the grain disappearances' rate in case **B**.

Spaepen, and P. Wollants did not observe large amounts of stable junctions of orders greater than 3 in their phase field simulations for boundary energy densities of Types I,II. Thus, the fact that our model operates with stable triple junctions but not with stable higher-order junctions does not seem to be very restrictive.

Remark 4. Figs. 14 and 15 imply that relation (33) predicts a slightly smaller boundary lengthening at small misorientations than it follows from Ref. [10, section3.3, Figures 8a,b,9a,b].

Remark 5. From Figs. 6-9, we see that the number- and lengthweighted MDFs tend to the limiting finite states when *N* decreases. For small *N*, the MDFs evolve slightly and mainly at low misorientations. This conforms with the simulation results introduced in Ref. [10, section 3.3, Figures 8(a,b), 9(a,b)] and [20, Figure 3(b)], where approximate steady states of the MDFs were indicated.

Remark 6. Figs. 16–19 show that when *N* decreases, the average boundary energy density $\overline{\gamma}(N)$ and the fraction $\kappa_{\text{he}}(N)$ of highenergy boundaries decrease, whereas the fraction $\kappa_{\text{le}}(N)$ of lowenergy boundaries increases. Such behavior persists up to some slight computational inaccuracies observed for small *N*, and it is a priori evident due to the energy reducing nature of grain growth. Furthermore, these quantities eventually tend to the corresponding approximate steady states.

Remark 7. According to Fig. 20 for case **B**, when *N* decreases, the ratio $\lambda(N)$ of the neighbor switchings' rate to the grain disappearances' rate also decreases and eventually reaches its approximate steady state.

8. Concluding remarks

In this paper, we proposed a novel kinetic modeling approach for estimating number- and length-weighted misorientation distribution functions in the particular technologically important case of polycrystalline thin films with fiber textures and with anisotropic grain boundary energy densities. The number-weighted misorientation distribution function was obtained numerically by solving a Cauchy problem for the derived first-order partial differential equation, in which the time variable was replaced with the varying total number of grain boundaries. The ratio of the lengthand number-weighted misorientation distribution functions was estimated by the use of a specific boundary lengthening model.

Despite some possible shortcomings of the developed framework, it was successfully tested by comparing the numerical results, which it gave for two nontrivial benchmark examples, with the corresponding large-scale simulation results presented in Refs. [10,20]. One auxiliary benchmark example was also considered. At the same time, our kinetic modeling approach based on the derived partial differential equation is a priori less computationally expensive than the generally recognized large-scale simulation techniques mentioned in the introduction.

Our approach has both theoretical as well as practical value. It can be used to elucidate fundamental mechanisms of texture evolution in two-dimensional polycrystalline systems.

As a possible future work, we can consider an extension of the developed approach by taking nucleations of new grains into account. The current framework belongs to a pure-death category and does not allow for any sort of grain birth. This may account for some deviations from experiments, where so-called defects are allowed to nucleate at certain rates; see, for instance, [21]. To extend our model by including a nucleation term and therefore by reformulating the kinetic equation seems to be worth investigating. Here one of the main difficulties lies in identifying a suitable nucleation regime.

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