







Review

Fiber Orientation Predictions—A Review of Existing Models

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Received: 11 May 2020; Accepted: 2 June 2020; Published: 8 June 2020



Abstract: Fiber reinforced polymers are key materials across different industries. The manufacturing processes of those materials have typically strong impact on their final microstructure, which at the same time controls the mechanical performance of the part. A reliable virtual engineering design of fiber-reinforced polymers requires therefore considering the simulation of the process-induced microstructure. One relevant microstructure descriptor in fiber-reinforced polymers is the fiber orientation. This work focuses on the modeling of the fiber orientation phenomenon and presents a historical review of the different modelling approaches. In this context, the article describes different macroscopic fiber orientation models such as the Folgar-Tucker, nematic, reduced strain closure (RSC), retarding principal rate (RPR), anisotropic rotary diffusion (ARD), principal anisotropic rotary diffusion (pARD), and Moldflow rotary diffusion (MRD) model. We discuss briefly about closure approximations, which are a common mathematical element of those macroscopic fiber orientation models. In the last section, we introduce some micro-scale numerical methods for simulating the fiber orientation phenomenon, such as the discrete element method (DEM), the smoothed particle hydrodynamics (SPH) method and the moving particle semi-implicit (MPS) method.

Keywords: fiber orientation; fiber reinforced thermoplastics; modeling

1. Introduction

Fiber reinforced polymers are key materials across different industries. For example short fiber reinforced thermoplastics are widely used in the automotive industry to reduce weight. The manufacturing processes of those materials have typically strong impact on their final microstructure, which at the same time controls the mechanical performance of the part. A reliable virtual engineering design of fiber-reinforced polymers therefore requires considering the simulation of the process-induced microstructure.

One relevant microstructure descriptor in fiber-reinforced polymers is the fiber orientation. This work focuses on the modeling of the fiber orientation phenomenon and presents a historical review of the different modeling approaches. In this context, the article describes different modeling approaches such as the addition of a scalar diffusion by Folgar and Tucker [1], the nematic potential approach [2], the modeling of a retarding rate in the reduced strain closure (RSC) [3] and the retarding principle rate (RPR) model [4], and lastly the anisotropic rotary diffusion approach (ARD) by Phelps et al. [5]. Additionally, reduced parameters models like the improved anisotropic rotary diffusion (iARD) [4], principal anisotropic rotary diffusion (pARD) [6], and Moldflow rotary diffusion (MRD) [7] will be introduced. The mentioned models are provided in commercial injection molding software such as Autodesk Moldflow[®], Moldex 3D[®], Sigmasoft[®] and Cadmould[®]. For example,

Autodesk Moldflow[®] provides the Folgar-Tucker, RSC and ARD-RSC model and Moldex 3D[®] the Folgar-Tucker, ARD, and iARD-RPR model.

Furthermore, we briefly discuss closure approximations, which are a common mathematical element of those macroscopic fiber orientation models. Simple closure approximation like the linear, quadratic, and hybrid closure [8,9] will be introduced. Then, we focus on exact closure approximations and fitted closure approximations. In the field of fitted closures we distinguish between orthotropic and invariant based closures.

Afterwards, we introduce micro-scale numerical methods for simulating the fiber orientation phenomenon, such as the discrete element method (DEM), smoothed particle hydrodynamics (SPH) and moving particle semi-implicit MPS. The focus will be on DEM based method and existing approaches will be looked at under the points of fiber discretization, imposed flow fields, fluid-fiber interaction, and fiber-fiber interaction.

The last section focuses on combining the advantages of both scales, for example through parameter fitting or machine learning approaches.

2. Fiber Orientation

The orientation of a single fiber can be characterized by a unit vector $\mathbf{p} \in \mathbb{S}^3 := \{\mathbf{p} \in \mathbb{R}^3 : \|\mathbf{p}\| = 1\}$ along the fiber axis. All $\mathbf{p} \in \mathbb{S}^3$ can be defined by two angles (ϕ, θ) (Figure 1)

$$p_1 = \sin \theta \cos \phi \tag{1}$$

$$p_2 = \sin \theta \sin \phi \tag{2}$$

$$p_3 = \cos \theta \tag{3}$$

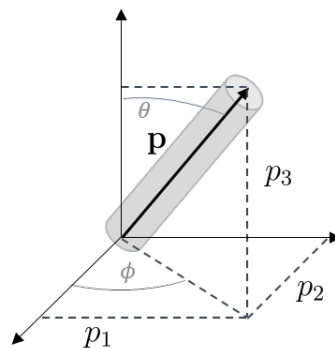


Figure 1. Orientation of a single rigid fiber \mathbf{p} .

To describe the orientation of many fibers statistical methods are useful. The probability density function (PDF) $\psi(\cdot)$ [1] is defined, such that $\psi(\mathbf{p}, t)d\mathbf{p}$ is the probability that a fiber is directed between \mathbf{p} and $\mathbf{p} + d\mathbf{p}$ at time t . The PDF has the following distinct properties.

$$\mathcal{B}(\psi) = [0, 1] \tag{4}$$

$$\oint \psi(\mathbf{p}, t)d\mathbf{p} = 1 \tag{5}$$

$$\psi(\mathbf{p}) = \psi(-\mathbf{p}) \tag{6}$$

$$\frac{D\psi}{Dt} = -\nabla_s \cdot (\psi\mathbf{p}). \tag{7}$$

Equation (6) is valid under the assumption that the fibers are cylindrical and have no preferred end. Equation (7) is called the continuity equation. The operator ∇_s represent the gradient operator on the surface of the unit sphere.

Since the PDF is defined on the unit sphere, computation is expensive and numerically difficult. For that reason, moments of the PDF are commonly used for computational efficiency [8]. The orientation tensors are defined by

$$a_{ij} = \int p_i p_j \psi(\mathbf{p}) d\mathbf{p} \tag{8}$$

$$a_{ijkl} = \int p_i p_j p_k p_l \psi(\mathbf{p}) d\mathbf{p}. \tag{9}$$

$$a_{i\dots n} = \int p_i \cdots p_n \psi(\mathbf{p}) d\mathbf{p}. \tag{10}$$

Since the PDF is even (Equation (6)) all odd-ordered orientation tensors are zero. To simplify notation we introduce $\mathbf{A} = a_{ij}$, $\mathbb{A} = a_{ijkl}$. The orientation tensors have important properties [8], which are described here for the second and fourth moment only. The second order tensor is symmetric and has a unit trace.

$$\mathbf{A}_{ij} = \mathbf{A}_{ji} \tag{11}$$

$$\text{tr}\mathbf{A} = 1. \tag{12}$$

The fourth order tensor is symmetric with respect to any pair of indices.

$$\mathbb{A}_{ijkl} = \mathbb{A}_{jikl} = \mathbb{A}_{ijlk} = \mathbb{A}_{kjl} = \mathbb{A}_{lkji} = \mathbb{A}_{ikjl} = \mathbb{A}_{ilkj} \tag{13}$$

and all information of the second order orientation tensor can be retrieved from the fourth order tensor

$$\mathbf{A}_{ij} = \mathbb{A}_{ijkk}. \tag{14}$$

The use of orientation tensors simplifies the computation because no discretization of the unit sphere is necessary, but it is impossible to distinguish between certain orientations. For example,

the orientation tensor for a bipolar and planar random orientation is identical with $\mathbf{A} = \begin{pmatrix} 0.5 & 0 & 0 \\ 0 & 0.5 & 0 \\ 0 & 0 & 0 \end{pmatrix}$.

An advantage is the objectivity, that means the equation is independent of the coordinate system.

3. Macroscopic Fiber Orientation Models

Macroscopic fiber orientation models are used to predict fiber orientation in parts. The models are integrated in commercial software such as Autodesk Moldflow[®], Moldex 3D[®], Sigmasoft[®] and Cadmould[®]. Additionally open source software such as OpenFoam can be used to implement the models.

3.1. Macroscopic Fiber Orientation Models in the Dilute Regime

The first description of the motion of a single fiber was developed by Jeffrey [10]. This description is based on the following assumptions: The fluid is Newtonian and has no turbulences ($\text{rot } \mathbf{u} = 0$). The particle is an ellipsoid and perfectly rigid, such that no bending or breaking occurs. The velocity field of the fluid is not influenced by the particle and there exists a perfect contact between the particle and the fluid. Under these assumptions the model is accurate up to order two. This has been proven in a more general way by Junk and Illner [11] under the same assumptions. Jeffrey's model has the form:

$$\dot{\mathbf{p}} = \mathbf{W} \cdot \mathbf{p} + \zeta(\mathbf{D} \cdot \mathbf{p} - \mathbf{D} : \mathbf{p}\mathbf{p}\mathbf{p}), \tag{15}$$

where $\mathbf{W} = \frac{1}{2}((\nabla \mathbf{u})^T - \nabla \mathbf{u})$ is the vorticity tensor, $\mathbf{D} = \frac{1}{2}((\nabla \mathbf{u})^T + \nabla \mathbf{u})$ the rate of strain tensor, $\zeta = \frac{r_e^2 - 1}{r_e^2 + 1}$ the particle shape function and ∇ the nabla operator. Applying it to injection molding

simulation, it has to be highlighted that the polymer melt is non-Newtonian and fibers are not perfectly rigid. In fact they can break and bend. Furthermore for highly filled polymers, the fluid is influenced by the fiber [2]. Coupling between fluid and fiber orientation will not be considered in this review.

3.2. Macroscopic Fiber Orientation Models in the Concentrated Regime

In the concentrated regime fiber interaction is dominant for fiber orientation. All macroscopic modeling approaches published up to today, accounting for interaction, are phenomenological.

Folgar and Tucker [1] added a diffusion term to account for fiber interaction to predict the orientation in semi-dilute and concentrated solutions. The model is valid under the following additional assumptions: The fibers are rigid cylinders, uniform in length and diameter. Moreover, they are sufficient large such that Brownian motion is negligible. The matrix is incompressible and sufficient viscous, such that particle inertia and buoyancy is negligible. The center of mass of the fibers are randomly distributed and no external forces or torques act on the suspension.

The fiber motion of a single fiber can then be described by

$$\dot{\mathbf{p}} = \mathbf{W} \cdot \mathbf{p} + \zeta(\mathbf{D} \cdot \mathbf{p} - \mathbf{D} : \mathbf{p}\mathbf{p}\mathbf{p}) + C_I \dot{\gamma} \cdot \nabla_s \psi, \quad (16)$$

where C_I describes the interaction coefficient and $\dot{\gamma} = \sqrt{2\mathbf{D} : \mathbf{D}}$ the scalar magnitude of the rate of strain tensor. The equation is, in contrast to Jeffrey's equation (15), not reversible. The interaction coefficient C_I is empirically determined and describes the rate of interaction. Setting $C_I = 0$ retrieves Jeffrey's equation (15). Interactions of fibers cause random orientation. The Fokker-Planck equation (17) [1] expresses the rate of change for the PDF, using the Folgar-Tucker equation for a single fiber (16) and the continuity equation (7). Fibers are modeled as independent, identically distributed, random variables with zero mean. Each interaction causes an orientation change in both fibers

$$\begin{aligned} \frac{D\psi}{Dt} &= -\nabla_s \cdot (\psi(\mathbf{W} \cdot \mathbf{p} + \zeta(\mathbf{D} \cdot \mathbf{p} - \mathbf{D} : \mathbf{p}\mathbf{p}\mathbf{p}) + C_I \dot{\gamma} \cdot \nabla_s \psi)) \\ &= -\nabla_s \cdot (\psi(\mathbf{W} \cdot \mathbf{p} + \zeta(\mathbf{D} \cdot \mathbf{p} - \mathbf{D} : \mathbf{p}\mathbf{p}\mathbf{p}))) + C_I \dot{\gamma} \cdot \nabla_s^2 \psi. \end{aligned} \quad (17)$$

The concept introduced by Folgar and Tucker offers advantages in the prediction of fiber orientation but is difficult to solve numerically and can only be solved with high computational effort. Based on the Fokker-Planck equation (17), Advani and Tucker [8] developed an equation for the rate of change of the second order orientation tensor

$$\frac{D\mathbf{A}}{Dt} = \dot{\mathbf{A}} = \dot{\mathbf{A}}^h + \dot{\mathbf{A}}^d \quad (18)$$

$$\dot{\mathbf{A}}^h = (\mathbf{W} \cdot \mathbf{A} - \mathbf{A} \cdot \mathbf{W}) + \zeta(\mathbf{D} \cdot \mathbf{A} + \mathbf{A} \cdot \mathbf{D} - 2\mathbb{A} : \mathbf{D}) \quad (19)$$

$$\dot{\mathbf{A}}^d = 2C_I \dot{\gamma}(\mathbf{I} - 3\mathbf{A}).. \quad (20)$$

The influence of the phenomenological parameter C_I is displayed in Figure 2. With decreasing diffusion (small C_I) fibers are more aligned.

Evolution equations of the second order orientation tensor \mathbf{A} contain the fourth order fiber orientation tensor \mathbb{A} . It is possible to derive an equation for the fourth order orientation tensor, but it will contain the sixth order orientation tensor. This leads to an infinite series of evolution equations. Therefore, it is common to truncate the series at the level of the second order orientation tensor. Since the fourth order orientation tensor is then not explicitly computed, a so-called closure approximation is necessary for calculating the fourth order fiber orientation tensor. In Section 3.3 different closure approximations will be introduced.

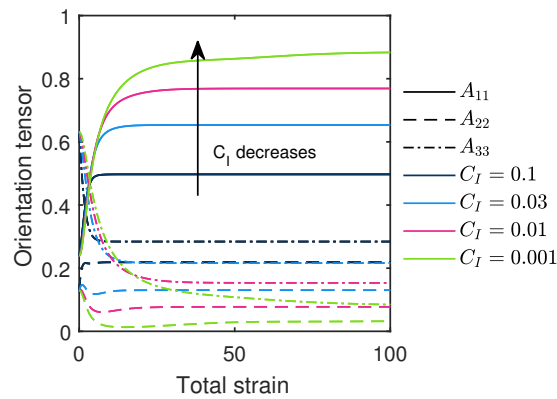


Figure 2. Influence of the phenomenological parameter C_I on fiber orientation evolution.

Three model enhancements have been made since the simulated fiber orientation shows deviation to fiber orientation determined experimentally.

To slow down the evolution speed, Huynh [12] introduced the strain reduction factor (SRF) $\frac{1}{\alpha}$ by multiplying equation (18) with $\alpha \in [0, 1]$

$$\dot{\mathbf{A}} = \alpha(\dot{\mathbf{A}}^h + \dot{\mathbf{A}}^d). \tag{21}$$

The SRF model violates the material objectivity, so Wang et al. [3] developed the reduced strain closure (RSC), based on the eigenvalue λ and eigenvector \mathbf{e} decomposition, that is, $\mathbf{A} = \sum_{i=1}^3 \lambda_i \mathbf{e}_i \mathbf{e}_i$ with orthonormal eigenvectors. The modified growth rate has the following form

$$\dot{\lambda}_i^{\text{RSC}} = \kappa \dot{\lambda}_i \tag{22}$$

$$\dot{\mathbf{e}}_i^{\text{RSC}} = \dot{\mathbf{e}}_i, \tag{23}$$

with the constant $\kappa \in [0, 1]$. In the differential equation form, this equals

$$\dot{\mathbf{A}} = \dot{\mathbf{A}}^{\text{RSC}} + \kappa \dot{\mathbf{A}}^d \tag{24}$$

$$\dot{\mathbf{A}}^{\text{RSC}} = \mathbf{W} \cdot \mathbf{A} - \mathbf{A} \cdot \mathbf{W} + \zeta [\mathbf{D} \cdot \mathbf{A} + \mathbf{A} \cdot \mathbf{D} - 2(\mathbb{A} + (1 - \kappa)(\mathbb{L} - \mathbb{M} : \mathbb{A})) : \mathbf{D}] \tag{25}$$

$$\mathbb{L} = \sum_{i=1}^3 \lambda_i \mathbf{e}_i \mathbf{e}_i \mathbf{e}_i \mathbf{e}_i \tag{26}$$

$$\mathbb{M} = \sum_{i=1}^3 \mathbf{e}_i \mathbf{e}_i \mathbf{e}_i \mathbf{e}_i. \tag{27}$$

Figure 3 displays the influence of the phenomenological constant κ with a smaller retarding rate (smaller κ) a slower orientation evolution is reached. The steady state is unchanged by κ . Tseng et al. [4] introduced at the same time the retarding principal rate (RPR), modifying the growth rate of the eigenvalue by

$$\dot{\lambda}_i^{\text{RPR}} = \kappa \dot{\lambda}_i + (1 - \kappa) \beta (\dot{\lambda}_i^2 + \dot{\lambda}_j \dot{\lambda}_k) \tag{28}$$

$$\dot{\mathbf{e}}_i^{\text{RPR}} = \dot{\mathbf{e}}_i, \tag{29}$$

with a fine-tuning parameter β . For $\beta = 0$ this equals the RSC model equation (22). The differential equation form (similar to (24) with PRR) can be developed based on equations (28) and (29).

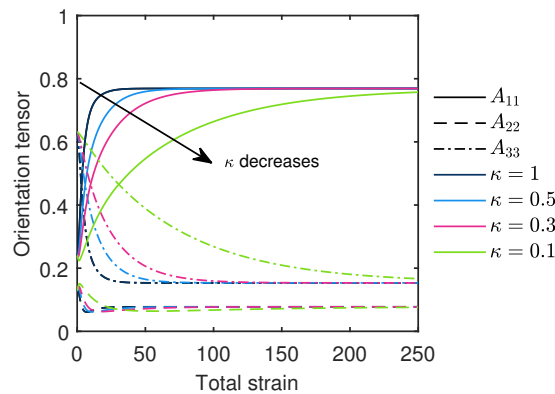


Figure 3. Influence of the phenomenological parameter κ on fiber orientation evolution. The diffusion constant is set to $C_I = 0.01$.

The second model approach, added by Phelps and Tucker [5], introduced an anisotropic rotary diffusion (ARD) term by replacing C_I with a rotary diffusion tensor \mathbf{C} . This allows spatially non-uniform rotary diffusion, which makes the rotary diffusion effect a function of the orientation state

$$\dot{\mathbf{A}} = \dot{\mathbf{A}}^h + \dot{\mathbf{A}}^{\text{ARD}} \tag{30}$$

$$\dot{\mathbf{A}}^{\text{ARD}} = \dot{\gamma} [2\mathbf{C} - 2\text{tr}(\mathbf{C})\mathbf{A} - 5(\mathbf{C} \cdot \mathbf{A} + \mathbf{A} \cdot \mathbf{C}) + 10\mathbf{A} : \mathbf{C}]. \tag{31}$$

The different modeling approaches for the anisotropic rotary diffusion are listed below

$$\mathbf{C} = \mathbf{C}(\mathbf{D}, \mathbf{A}) = b_1\mathbf{I} + b_2\mathbf{A} + b_3\mathbf{A}^2 + \frac{b_4}{\dot{\gamma}}\mathbf{D} + \frac{b_5}{\dot{\gamma}^2}\mathbf{D}^2. \tag{32}$$

ARD model [5] with constants $b_i, i = 1, \dots, 5$

$$\mathbf{C} = C_I \left(\mathbf{I} - 4C_M \frac{\mathbf{D}^2}{\dot{\gamma}} \right) \tag{33}$$

iARD model [4] with constants C_I, C_M

$$\mathbf{C} = C_I \mathbf{R}_A \begin{pmatrix} D_1 & 0 & 0 \\ 0 & D_2 & 0 \\ 0 & 0 & D_3 \end{pmatrix} \mathbf{R}_A^\top \tag{34}$$

pARD model [6] with constants C_I, D_1, D_2, D_3 .

The eigenmatrix \mathbf{R}_A is defined by

$$\mathbf{A} = \mathbf{R}_A \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix} \mathbf{R}_A^\top. \tag{35}$$

Tseng et al. [6] chose

$$D_1 = 1, D_2 = c, D_3 = 1 - c \tag{36}$$

to reduce the needed amount of parameters. This implies that the rotary diffusion factor in the first principal fiber orientation direction is defined by C_I . The second principal fiber orientation direction is scaled with $c \cdot C_I$ and in the third principal fiber orientation direction the smallest rotary diffusion factor with $(1 - c)C_I$ is applied. The influence of D_2 is displayed in Figure 4.

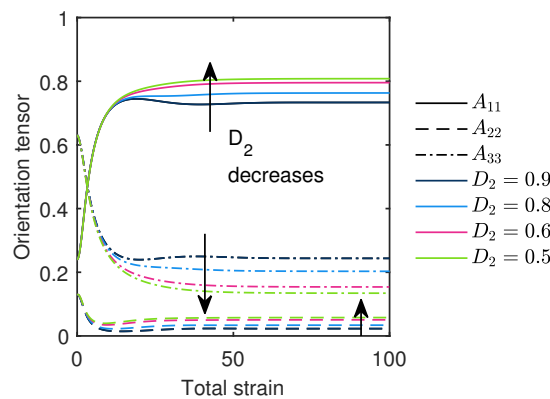


Figure 4. Influence of the phenomenological parameter D_2 on fiber orientation evolution. The diffusion is set to $C_I = 0.01$.

A slightly different approach to add anisotropic rotary diffusion is defined in the MRD model [7]. The rotary diffusion is modeled according to equation (34), but not the full ARD equation (31) is used. The reduced form is given by

$$\dot{\mathbf{A}} = \dot{\mathbf{A}}^h + \dot{\mathbf{A}}^{mARD} \tag{37}$$

$$\dot{\mathbf{A}}^{mARD} = 2\dot{\gamma}(\mathbf{C} - \text{tr}(\mathbf{C})\mathbf{A}). \tag{38}$$

Recently, Favaloro and Tucker [13] published a framework to compare anisotropic rotary diffusion approaches and compared the mentioned approaches in shear and elongation flows. They added a suggestion for a more general but stable model by using equation (34) but making D_i a function of λ_i or using equation (32) with $b_4 = b_5 = 0$ and $b_i, i = 1, 2, 3$ as scalar functions of $\text{tr}\mathbf{A}^2$ and $\text{tr}\mathbf{A}^3$. Latz et al. [2] added a nematic potential to the Folgar-Tucker model to account for excluded volume effects. So far, there is no clear dependency of C_I on the volume fraction and aspect ratio of the fillers. Different experiments in different regimes even showed contradictory results [14–16]. Latz et al. [2] stated that the excluded volume mechanism may be a possible explanation for the observed effects. The integration of a second constant U_0 could decouple these effects and perhaps give a clear dependence on physical descriptors. The model is defined by

$$\dot{\mathbf{A}} = \dot{\mathbf{A}}^h + \dot{\mathbf{A}}^{nem} \tag{39}$$

$$\dot{\mathbf{A}}^{nem} = (C_I(\mathbf{I} - 3\mathbf{A}) + U_0(\mathbf{A} \cdot \mathbf{A} - \mathbf{A} : \mathbb{A}))\dot{\gamma}, \tag{40}$$

with one additional constant U_0 .

In injection molding simulation combination of the methods are used. Combining Equations (24) and (31) leads to the following (p)ARD-RSC model

$$\dot{\mathbf{A}} = \dot{\mathbf{A}}^{RSC} + \dot{\mathbf{A}}^{(p)ARD-RSC} \tag{41}$$

$$\begin{aligned} \dot{\mathbf{A}}^{(p)ARD-RSC} = & \dot{\gamma}[2[\mathbf{C} - (1 - \kappa)\mathbb{M} : \mathbf{C}] - 2\kappa(\text{tr}\mathbf{C})\mathbf{A} - 5(\mathbf{C} \cdot \mathbf{A} + \mathbf{A} \cdot \mathbf{C}) \\ & + 10[\mathbb{A} - (1 - \kappa)(\mathbb{L} - \mathbb{M} : \mathbb{A})] : \mathbf{C}]. \end{aligned} \tag{42}$$

3.3. Closure Approximations

Since the fourth order orientation tensor is used in all models, closure approximations are needed. Any fourth order tensors with the symmetric properties stated in equation (13) can be represented by a 6×6 matrix with at most 36 independent entries [17]

$$\underline{\mathbb{A}} = \begin{pmatrix} \mathbb{A}_{11} & \mathbb{A}_{12} & \mathbb{A}_{13} & \mathbb{A}_{14} & \mathbb{A}_{15} & \mathbb{A}_{16} \\ \mathbb{A}_{21} & \mathbb{A}_{22} & \mathbb{A}_{23} & \mathbb{A}_{24} & \mathbb{A}_{25} & \mathbb{A}_{26} \\ \mathbb{A}_{31} & \mathbb{A}_{32} & \mathbb{A}_{33} & \mathbb{A}_{34} & \mathbb{A}_{35} & \mathbb{A}_{36} \\ \mathbb{A}_{41} & \mathbb{A}_{42} & \mathbb{A}_{43} & \mathbb{A}_{44} & \mathbb{A}_{45} & \mathbb{A}_{46} \\ \mathbb{A}_{51} & \mathbb{A}_{52} & \mathbb{A}_{53} & \mathbb{A}_{54} & \mathbb{A}_{55} & \mathbb{A}_{56} \\ \mathbb{A}_{61} & \mathbb{A}_{62} & \mathbb{A}_{63} & \mathbb{A}_{64} & \mathbb{A}_{65} & \mathbb{A}_{66} \end{pmatrix}, \tag{43}$$

where $\mathbb{A}_{\alpha\beta} = \mathbb{A}_{ijkl}$ for $\alpha, \beta \in \{1, 2, 3, 4, 5, 6\}$ the indices α and β represent an index pair ij or kl in the following way:

$$\alpha, \beta : \begin{cases} 1 \rightarrow 11 \\ 2 \rightarrow 22 \\ 3 \rightarrow 33 \\ 4 \rightarrow 23 \\ 5 \rightarrow 31 \\ 6 \rightarrow 12. \end{cases} \tag{44}$$

The first simple closure approaches have been introduced by Advani and Tucker [8]. The linear approach [8] is a summation of all products of a_{ij} and δ_{ij} . After applying symmetry and normalization condition the following linear approximation occurs

$$\begin{aligned} \mathbb{A}_{ijkl}^{lin} &= -\frac{1}{35}(\delta_{ij}\delta_{kl} + \delta_{ik}\delta_{jl} + \delta_{il}\delta_{jk}) \\ &+ \frac{1}{7}(\mathbf{A}_{ij}\delta_{kl} + \mathbf{A}_{ik}\delta_{jl} + \mathbf{A}_{il}\delta_{jk} + \mathbf{A}_{kl}\delta_{ij} + \mathbf{A}_{jl}\delta_{ik} + \mathbf{A}_{jk}\delta_{il}). \end{aligned} \tag{45}$$

The quadratic closure [8] omits all linear terms

$$\mathbb{A}_{ijkl}^{quad} = \mathbf{A}_{ij}\mathbf{A}_{kl}. \tag{46}$$

The quadratic closure does not preserve the symmetry of the fourth order tensor (equation (8)), but applying it to equation (19), it does preserve the symmetry of the second order tensor.

It is also possible to combine both approaches. This leads to the hybrid closure [8]

$$\mathbb{A}_{ijkl}^{hyb} = (1 - f)\mathbb{A}_{ijkl}^{lin} + f\mathbb{A}_{ijkl}^{quad} \tag{47}$$

$$f = \frac{3}{2}\mathbf{A}_{ij}\mathbf{A}_{ji} - \frac{1}{2}. \tag{48}$$

Another approach to determine the function for the hybrid closure was determined by Advani and Tucker [9]

$$f = 1 - 27 \cdot \det \mathbf{A}. \tag{49}$$

A more advanced approach is the natural closure (NAT). It is based on the relationship $\mathbb{A} = f(\mathbf{A})$ when no diffusion occurs ($C_I = 0$).

Verley and Dupret [18] stated that there exists an exact closure in the case that the orientation is at one time isotropic and used a numerical approximation in the 3-D case to obtain manageable computational cost.

Montgomery-Smith et al. [19] determined the exact formulation in Cartesian coordinates on the sphere using the Carlson form of elliptic integrals. They used the analytic solution of the Jeffrey equation presented by Dinh and Armstrong [20]. The approach is based on the assumption of isotropic orientation at $t = 0$. Then the second order orientation tensor can be expressed by

$$\mathbf{A} = \int_S \frac{\mathbf{p}\mathbf{p}}{4\pi(\mathbf{B} : \mathbf{p}\mathbf{p})^{\frac{3}{2}}} d\mathbf{p} \tag{50}$$

$$\mathbf{B} = -\mathbf{B} \cdot (\mathbf{W} + \zeta\mathbf{D}) - (-\mathbf{W} + \zeta\mathbf{D}) \cdot \mathbf{B}, \mathbf{B} = \mathbf{I} \text{ at } t = 0. \tag{51}$$

The method presented uses high computational effort, so Montgomery-Smith et al. [19] introduced the fast exact closure (FEC). The FEC introduced a computationally efficient way to compute the closure. Instead of computing \mathbf{B} by inverting the integral, equations (18) and (51) are solved simultaneously. If the initial data is not isotropic, \mathbf{B} has to be computed for the initial condition.

Later Montgomery-Smith et al. [21] also include the anisotropic rotary diffusion of Phelps and Tucker [5]. In this case ($\mathbf{C} \neq 0$) the closure is not exact. The key idea is to introduce a matrix \mathbb{B} and define two conversion tensors \mathbb{C} and \mathbb{D} such that

$$\frac{D\mathbf{A}}{Dt} = -\mathbb{C} : \frac{D\mathbf{B}}{Dt}, \frac{D\mathbf{B}}{Dt} = -\mathbb{D} : \frac{D\mathbf{A}}{Dt}. \tag{52}$$

hold true. The ordinary differential equations (ODEs) are solved simultaneously, which can be computed very efficiently. The special form used for this approach is

$$\frac{D\mathbf{A}}{Dt} = -\mathbb{C} : F(\mathbf{B}) + G(\mathbf{A}), \frac{D\mathbf{B}}{Dt} = F(\mathbf{B}) - \mathbb{D} : G(\mathbf{A}). \tag{53}$$

They showed that it can also be applied to the reduced strain closure in equation (24) and proved that the solution stays physical for example, that \mathbf{A} stays positive definite with $\text{tr } \mathbf{A} = 1$.

A large family of closure are fitted closures. Chung and Kwon [22] stated two ways to develop fitted closure. The method depends on the coordinate system, global or eigenspace, and the representation of the approximation, invariants or eigenvalues, of the orientation tensor. The orthotropic closures use the eigenspace coordinate system and eigenvalues as representatives. They are the most used family of closures. Cintra and Tucker [23] stated that an objective closure has to be orthotropic. Orthotropic means that the principal axes must match those of the second order tensor and each principal fourth order component is a function of just two principal values of the second order tensor.

The second order orientation tensor can be transformed in the eigenraum representation

$$\hat{\mathbf{A}} = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix} = \mathbf{R}_A^T \mathbf{A} \mathbf{R}_A, \tag{54}$$

with non-negative eigenvalues $\lambda_1 \geq \lambda_2 \geq \lambda_3 \geq 0$ and $\lambda_1 + \lambda_2 + \lambda_3 = 1$. Figure 5 shows the reference triangle for orthotropic closures.

Kuzmin [17] showed that regarding all symmetries of the fourth order orientation tensor the orthotropic representation has the form

$$\hat{\underline{\underline{\mathbf{A}}}} = \begin{pmatrix} \hat{\underline{\underline{A}}}_{11} & \hat{\underline{\underline{A}}}_{12} & \hat{\underline{\underline{A}}}_{13} & 0 & 0 & 0 \\ \hat{\underline{\underline{A}}}_{12} & \hat{\underline{\underline{A}}}_{22} & \hat{\underline{\underline{A}}}_{23} & 0 & 0 & 0 \\ \hat{\underline{\underline{A}}}_{13} & \hat{\underline{\underline{A}}}_{23} & \hat{\underline{\underline{A}}}_{33} & 0 & 0 & 0 \\ 0 & 0 & 0 & \hat{\underline{\underline{A}}}_{44} & 0 & 0 \\ 0 & 0 & 0 & 0 & \hat{\underline{\underline{A}}}_{55} & 0 \\ 0 & 0 & 0 & 0 & 0 & \hat{\underline{\underline{A}}}_{66} \end{pmatrix}. \tag{55}$$

It has been proven by Cintra and Tucker [23] and Kuzmin [17] that the fourth order orientation tensor can be represented by three independent components. This is due to the orthotropic properties, the symmetries of the second order tensor, the symmetry with respect to any pair of indices for the fourth order tensor and the normalization condition. The three components can be expressed as functions of the two largest eigenvalues of the second order tensor

$$\hat{\underline{\underline{A}}}_{11} = f_1(\lambda_1, \lambda_2) \tag{56}$$

$$\hat{\underline{\underline{A}}}_{22} = f_2(\lambda_1, \lambda_2) \tag{57}$$

$$\hat{\underline{\underline{A}}}_{33} = f_3(\lambda_1, \lambda_2). \tag{58}$$

The remaining entries are defined by symmetry and normalization condition

$$\hat{\underline{\underline{A}}}_{12} = \hat{\underline{\underline{A}}}_{66} \tag{59}$$

$$\hat{\underline{\underline{A}}}_{23} = \hat{\underline{\underline{A}}}_{44} \tag{60}$$

$$\hat{\underline{\underline{A}}}_{13} = \hat{\underline{\underline{A}}}_{55} \tag{61}$$

$$\hat{\underline{\underline{A}}}_{55} + \hat{\underline{\underline{A}}}_{66} = \lambda_1 - \hat{\underline{\underline{A}}}_{11} \tag{62}$$

$$\hat{\underline{\underline{A}}}_{44} + \hat{\underline{\underline{A}}}_{66} = \lambda_2 - \hat{\underline{\underline{A}}}_{22} \tag{63}$$

$$\hat{\underline{\underline{A}}}_{44} + \hat{\underline{\underline{A}}}_{55} = \lambda_3 - \hat{\underline{\underline{A}}}_{33}. \tag{64}$$

Cintra and Tucker [23] used the eigenspace coordinate system and eigenvalues as representatives to fit the closure (EBOF). Complete second order polynomials were used to approximate the components (ORF). The orthotropic fitted (ORF) closure and the NAT closure are based on identical assumptions. In conclusion they can be transferred to each other. In fact they are mathematically equivalent, but the ORF closure is numerically more stable for repeated eigenvalues [19] The three remaining components are fitted by

$$\hat{\underline{\underline{A}}}_{ii} = f_i(\lambda_1, \lambda_2) = C_i^1 + C_i^2 \lambda_1 + C_i^3 \lambda_1^2 + C_i^4 \lambda_2 + C_i^5 \lambda_2^2 + C_i^6 \lambda_1 \lambda_2 \tag{65}$$

for for $i = 1, 2, 3$.

The components are fitted with different flow types (simple shear, two shearing/stretching flows, uniaxial elongation, biaxial elongation) using a least squares routine. The closure approximation shows good agreement with the distribution function and the NAT and better performance than the hybrid closure. For small C_I the ORF closure shows oscillating behavior.

Chung and Kwon [24] improved the ORF closure to overcome the oscillation for small C_I values and introduced the orthotropic fitted closure approximation for wide interaction coefficients (OWE) and orthotropic fitted closure approximation for wide interaction coefficients with third order polynomial approximations (OWE3) closure. The OWE closure is fitted with two additional flow fields (shear/planar elongation, balanced shear/biaxial elongation) to cover the orientation triangle \hat{K} (Figure 5) more closely. Consequently the only difference between the ORF and OWE closure are the values of the fitted parameters. The OWE3 closure uses the same flows for the parameter fit but approximates the coefficients values by a third order polynomial expression

$$\begin{aligned} \hat{\underline{\underline{A}}}_{ii} = f_i(\lambda_1, \lambda_2) = & C_i^1 + C_i^2 \lambda_1 + C_i^3 \lambda_1^2 + C_i^4 \lambda_2 + C_i^5 \lambda_2^2 + C_i^6 \lambda_1 \lambda_2 \\ & + C_i^7 \lambda_1^2 \lambda_2 + C_i^8 \lambda_1 \lambda_2^2 + C_i^9 \lambda_1^3 + C_i^{10} \lambda_2^3 \end{aligned} \tag{66}$$

for for $i = 1, 2, 3$. The closures show stable behavior for a wide range of C_I but tend to over predict the orientation.

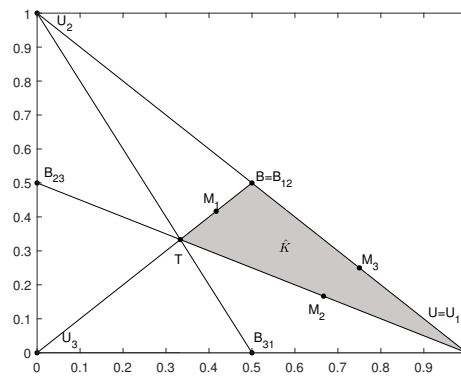


Figure 5. Reference triangle for orthotropic closure.

Kuzmin [17] introduced a mathematical concept to develop orthotropic closures. A concept for planar orientations was developed and extended to the 3D case. In this work only the 3D case is explained, for the planar case refer to Reference [17]. The linear and smooth closures were stated in the orthotropic state [17]. An orthotropic version of the quadratic closure was developed

$$\hat{\underline{\underline{A}}}_{11} = f_1(\lambda_1, \lambda_2) = \lambda_1^2 \tag{67}$$

$$\hat{\underline{\underline{A}}}_{22} = f_2(\lambda_1, \lambda_2) = \lambda_2^2 \tag{68}$$

$$\hat{\underline{\underline{A}}}_{33} = f_3(\lambda_1, \lambda_2) = (1 - \lambda_1 - \lambda_2)^2. \tag{69}$$

Since there does not exist an analytical form of the standard NAT, natural closures based on extended quadratic and piecewise linear interpolation have been developed.

The extended quadratic fit is fitted on cubic polynomials of the form

$$\begin{aligned} \hat{\underline{\underline{A}}}_{ii} = f_i(\lambda_1, \lambda_2) = & C_i^1 + C_i^2 \lambda_1 + C_i^3 \lambda_1^2 + C_i^4 \lambda_2 + C_i^5 \lambda_2^2 + C_i^6 \lambda_1 \lambda_2 \\ & + C_i^7 \lambda_1 \lambda_2 (1 - \lambda_1 - \lambda_2). \end{aligned} \tag{70}$$

The data points $U_i, B_{ij}, i = 1, 2, 3, j = i + 1$ for $i = 1, 2, j = 1$ for $i = 3$ and T are used, using the planar natural closure and the triaxial orientation state at T . For extrapolation the extended quadratic finite element method is used.

For the exact midpoint fit quadratic interpolation polynomials are used

$$\hat{\underline{\underline{A}}}_{ii} = f_i(\lambda_1, \lambda_2) = C_i^1 + C_i^2 \lambda_1 + C_i^3 \lambda_1^2 + C_i^4 \lambda_2 + C_i^5 \lambda_2^2 + C_i^6 \lambda_1 \lambda_2. \tag{71}$$

The values at points U, B, T and the midpoints M_1, M_2, M_3 are validated using the exact closure.

In contrast to all eigenspace based closures, Reference [22] introduced the invariants-based optimal fitted (IBOF) closure in the global coordinate system with invariants as representatives. This closure is computational more efficient than the orthotropic closures.

4. Microscopic Fiber Simulation

Simulation methods on the microscopic level can be used to simulate fiber movement for discretized fibers. In contrast to the phenomenological macroscopic fiber orientation models, modeling approaches on the microscopic scale enable the approximation of the physical behavior more accurately. However, they are computationally very expensive and are, in most cases, not suitable for the simulation of actual parts. Mostly, they are used on reference elements to investigate physical effects and quantities. Different modeling approaches can be used on the micro level. They can coarsely be divided in particle based methods, where the polymer matrix and the fibers are treated as particles for example the smoothed particle hydrodynamics (SPH) and moving particle semi-implicit

(MPS) method, or element based methods, which treat fibers as particles and the matrix is treated as a continuous media.

One big advantage of particle based method, is the computationally cheap coupling between fluid motion and fibers in two ways. Yashiro et al. [25,26] developed a method based on MPS to predict fiber movement in injection molding. They simulated complex flow fields and investigated orientation in a T-shaped bifurcation.

The SPH method is also applied to polymer composites. He et al. [27] simulated the 3D injection molding process for short-fiber reinforced polymers; Bertevras et al. [28] simulated the 3D printing process of fiber-reinforced polymer and Yamagata and Ichimiya [29] used the method to simulate the solidification process for injection molded short-fiber reinforced parts.

The SPH method can also be combined with the discrete element method (DEM) [30], or the element bending group (EBG) method [31,32]. The fluid is then model by the SPH method and the fibers by the respective method.

Element based methods are mostly developed on the principle of the DEM. In contrast to particle based methods, which are per se two-way coupled, DEM based simulations are often solved one-way coupled. The backcoupling from the fiber motion on the fluid is computationally expensive, but can be integrated in a coarse or refined way. The resulting fluid equation can be solved by multiple approaches such as the direct numerical simulation (DNS), the Lattice-Boltzmann or the particle finite element analysis (pFEA). In the DEM, fibers are considered as particles and the movement of each particle is calculated by the solution of the force and torque balance acting on each particle. Fibers are either discretized as chain of spheres, prolate spheroids or chain of rods. The forces acting on a fiber are

- hydrodynamic forces
- fiber fiber interaction forces
- elastic and bending forces (intra fiber forces)

Hydrodynamic forces are exerted from the fluid on the fiber. The fluid motion can be considered undisturbed by the fiber motion or disturbed. In the second case a backcoupling is necessary. The interaction forces can be divided in two cases: long-range hydrodynamic interaction and short range interaction. The short range interactions can then be divided in three regimes: short range lubrication forces, transition and mechanical contact. Fibers can be modeled flexible by using chains of beads or rods connected by joints. Breaking can also be incorporated at the defined joints. The modeling of the forces, the discretization of the fibers and the fluid motion varies between the different approaches and the following Table 1 gives an overview of published approaches, without claim for completeness. Only inertia free models are considered. Examples for models incorporating inertia are References [33–35].

Table 1. Literature overview of element based simulations for fiber reinforced polymers in chronological order.

	Discretization of Fibers	Flow Fields	Fluid-Fiber Interaction	Fiber-Fiber Interaction	Flexibility	Regarded Quantities
Yamamoto and Matsuoka 1993 [36]	chain of beads	shear	one-way coupled	-	flexible	single fiber movement
Yamamoto and Matsuoka 1994 [37]	chain of beads	shear	one-way coupled	-	flexible	viscosity of dilute solutions
Yamane et al., 1994 [38]	rods	shear	one-way coupled	lubrication	-	semi dilute suspensions, orientation evolution, diffusion constant, shear viscosity
Yamane et al., 1995 [39]	rods	shear	one-way coupled	lubrication	-	semi dilute suspensions, bounded and unbounded system
Yamamoto and Matsuoka 1995 [40]	chain of beads	shear	one-way coupled	lubrication	flexible	concentrated suspension, viscosity, stresses
Thomasset et al., 1997 [41]	rigid rods	varies flow fields	one-way coupled	lubrication, mechanical and hydrodynamical contact, no friction	-	2D, effects of fiber motion and orientation
Sundararajakumar and Koch 1997 [42]	rods	shear	one-way coupled	lubrication, mechanical and hydrodynamical contact, no friction	-	dilute: hydrodynamical contact most important, approaching higher concentration fiber contact
Skjetne et al., 1997 [43]	prolate spheroids	shear	one-way coupled	-	flexible, rigid	single fiber movement
Ross and Klingenberg 1997 [44]	prolate spheroids	shear	one-way coupled	repulsive interactions	flexible and rigid	single fiber movement, viscosity
Fan et al., 1998 [45]	rods	shear	one-way coupled	lubrication, no friction, long range hydrodynamic interactions by slender body theory	-	orientation, viscosity, stresses, all regimes
Harlen et al., 1999 [46]	rods	no imposed flow	-	mechanical contact, friction, long range hydrodynamic interactions by slender body theory	-	sphere settling through suspension of neutrally buoyant fibers, fiber contact has significant influence

Table 1. Cont.

	Discretization of Fibers	Flow Fields	Fluid-Fiber Interaction	Fiber-Fiber Interaction	Flexibility	Regarded Quantities
Phan-Thien et al., [14]	rods	shear	one-way coupled	lubrication, no friction, long range hydrodynamic interactions by slender body theory	-	FT constant, dilute and semi dilute
Joung et al., 2001 [47]	chain of spherical beads	shear and extensional flows	one-way coupled	lubrication, preventing from overlapping, long range hydrodynamic interactions	rigid, flexible	viscosity, orientation
Joung et al., 2002 [48]	chain of spherical beads	shear and complex flows	one-way coupled	lubrication, preventing from overlapping, long range hydrodynamic interactions	rigid, curved	viscosity for curved fibers
Joung et al., 2003 [49]	chain of spherical beads	shear and complex flows	one-way coupled	lubrication, preventing from overlapping, long range hydrodynamic interactions	flexible	Jeffrey orbits for rigid and flexible fibers, relationship between fiber stiffness, and bulk viscosity, arbitrary particle shapes, dilute regime effects of shape, friction, aspect ratio and stiffness, yield stress, rheology in flocculated systems
Switzer and Klingenberg 2003 [50]	chain of rods	shear	one-way coupled	mechanical interaction, friction	flexible	2D, effects of shear rate on flow behavior and micro structure, shear-induced self diffusion
Kromkamp et al., 2005 [51]	rods	shear	coupled, Lattice Boltzmann for fluid forces, particles as boundary surfaces	lubrication correction, mechanical interaction, no friction	-	orientation, viscosity, stresses, up to $\Phi = 11.5$
Ausias et al., 2006 [52]	rigid prolate spheroids	shear	one-way coupled	lubrication and interaction in normal direction, no friction, no long range interactions	-	optimal rod length for high accuracy and efficient calculation
Wang et al., 2006 [53]	rod-chain	shear	one-way coupled		flexible, rigid	

Table 1. Cont.

	Discretization of Fibers	Flow Fields	Fluid-Fiber Interaction	Fiber-Fiber Interaction	Flexibility	Regarded Quantities
Lindström and Uesaka 2007 [54]	rod-chain model	shear	coarse two-way coupling	lubrication and interaction in normal direction, friction	flexible	Jeffrey orbits, curvature, regimes of motions for flexible fibers
Lindström and Uesaka 2008 [55]	rod-chain model	shear	coarse two-way coupling	lubrication and interaction in normal direction, friction	flexible	orientation, viscosity, dilute and semidilute regime, rheological properties
Lindström and Uesaka 2009 [56]	rod-chain model	shear	coarse two-way coupling	lubrication and interaction in normal direction, friction	flexible	rheological properties
Yamanoi and Maia 2010 [57]	chain of beads	shear	one-way coupling	lubrication, mechanical contact, long range hydrodynamic interactions	-	rheological properties, orientation
Yamanoi et al., 2010 [58]	chain of beads	shear	one-way coupling	lubrication, mechanical contact, long range hydrodynamic interactions	flexible	nylon fiber, rheological properties, orientation, effect of flexibility
Yamanoi and Maia 2010 [59]	chain of beads	uniaxial elongation flow	one-way coupling	lubrication, mechanical contact, long range hydrodynamic interactions	flexible	rheological properties, orientation, orientation tensor independent of aspect ratio, volume fraction
Yamanoi and Maia 2011 [60]	chain of beads	shear	two way coupling	single fiber	rigid and flexible	hydrodynamic interaction in single fiber movement in shear
Andrić et al., 2013 [61]	rod-chain model	turbulent flow	two way coupling, DNS for fluid motion	single fiber	rigid and flexible	fiber-flow interaction for a single fiber
Andrić et al., 2014 [62]	rod-chain model	shear	two way coupling, DNS for fluid motion	-	rigid and flexible	dilute solution, no interaction, rheological properties, orbit drifts
Do-Quang et al., 2014 [63]	rod-chain model	turbulent flow	two way coupling, entropy lattice Boltzmann for fluid, external boundary force method	lubrication and mechanical contact	rigid	cellulose fibers in water, accumulation effects

Table 1. Cont.

	Discretization of Fibers	Flow Fields	Fluid-Fiber Interaction	Fiber-Fiber Interaction	Flexibility	Regarded Quantities
Mezher et al., 2015 [64]	prolate spheroids	shear	one-way coupling	lubrication and interaction in normal direction, no friction, no long range hydrodynamics	flexible	concentrated, orientation, normalized stresses, interactions, elastic energy
Mezher et al., 2016 [65]	prolate spheroids	shear	one-way coupling	lubrication and interaction in normal direction, no friction, no long range hydrodynamics	flexible	concentrated $\Phi = 7 - 18.2$, orientation, diffusion constants, confinement effects
Wang et al., 2016 [66]	rod-chain model	shear	one-way coupled	-	flexible	new rod chain model, optimal rod length
Perez et al., 2016 [67]	rod	shear	one-way coupling	only wall interaction	-	dilute, confinement effects
Sasayama and Inagaki 2017 [68]	simplified bead-chain model	shear	one-way	mechanical, lubrication	flexible	simplified bead-chain model for hydrodynamic calculations
Kuhn et al., 2017 [69]	rod-chain model	complex flow fields	one-way coupling	mechanical, friction, no long-range hydrodynamic	flexible	fiber matrix separation in compression molding, LFRT
Kuhn et al., 2018 [70]	rod-chain model	complex flow fields	one-way coupling	mechanical, friction, no long-range hydrodynamic	flexible	rib filling
Meirson and Hrymak 2018 [71]	rod-chain model	squeeze	one-way	-	flexible	2D, fiber orientation and deformation
Wu et al., 2018 [30]	bounded spheres	complex flow	2-way coupled, SPH for fluid motion	linear contact	-	2D, fiber orientation, accumulation during injection molding
Sasayama and Inagaki 2019 [72]	efficient bead-chain model	shear	one-way	mechanical, lubrication, friction	flexible	efficient bead-chain model for hydrodynamic calculations
Sasayama et al., 2019 [73]	efficient bead-chain mode	shear	one-way	mechanical, lubrication, friction	flexible, breakage	fiber breakage
Laurentcin et al., 2019 [74]	sphero-cylinder	squeeze flow (lubricated, non-lubricated)	one-way	-	rigid	non-Newtonian fluid, dilute regime, comparison between numerical analytical and experimental results

The listed approaches show, that depending on the evaluated quantities, the matrix and fiber material and the volume fraction, different modeling approaches show more promising results. In case of rigid fibers, long range hydrodynamic interactions show the highest influence in semidilute solution, whereas mechanical interaction gets dominant in the concentrated regime and long-range hydrodynamic interaction can be neglected [42,46,57]. The backcoupling has neglectable influence on rigid single fiber movement [60,75]. Once flexibility of the fiber is higher the backcoupling cannot be neglected [60]. Due to the high computational effort it has not been used in the concentrated regime. If there is a significant influence in the concentrated regime can not be answered.

Based on the movement of single fibers, orientation evolution curves can be calculated. With the discrete position of each fiber the second order orientation tensor can be calculated in each time step by

$$A_{ij}(t) = \sum_{n=1}^N \frac{p_{n,j}(t)p_{n,i}(t)}{N}, \quad (72)$$

where N denotes the number of fibers, t the actual time, and $p_n(t)$ the position of fiber n at time t .

5. Using Microscopic Models for an Enhanced Prediction on the Macroscopic Scale

A combination of two scales can enhance the prediction on the macroscopic scale, while being computationally cheap. A straight forward approach is to use the microscopic simulation for parameter definition of existing macroscopic model. This has been done by many authors, for example Reference [65,75]. Microscopic fiber orientation evolution results can also lead to new macroscopic models [67]. A different approach is to create orientation data with a microscopic simulation and use a machine learning based approach on the macroscopic level [76].

6. Summary

The fiber orientation phenomenon can be simulated either on a macro- or micro-scale by using current numerical techniques and state-of-the-art modelling approaches. Macroscopic fiber orientation models are computationally efficient and are the preferred solution for estimating fiber orientation in large industrial applications. The limitation of those models is their phenomenological nature and the dependency on a set of fitting parameters. The challenge is typically the correct choice of parameters for a specific material. On the other hand, microscopic fiber models have a larger physical basis, but since fibers are explicitly discretized they are not suitable for large-scale simulations, because they are computationally expensive. A combination of the models from the two scales can relieve the individual shortcomings and provides an interesting numerical solution for the virtual engineering design of fiber-reinforced polymer parts. One possibility is the prediction of optimal macroscopic parameters by a microscopic simulation, or the derivation of a macroscopic data-driven model using microscopic simulation results as input data.

Author Contributions: Conceptualization, S.K.K.; Writing—Original Draft Preparation, S.K.K.; Writing—Review and Editing, T.O., A.K. and C.C.; Visualization, S.K.K.; Supervision, T.O., A.K. and C.C. All authors have read and agreed to the published version of the manuscript.

Funding: This research received no external funding.

Conflicts of Interest: The authors declare no conflict of interest.

Abbreviations

The following abbreviations are used in this manuscript:

PDF	probability density function
ODE	ordinary differential equation
FT	Folgar-Tucker

nem	nematic
SRF	strain reduction factor
RSC	reduced strain closure
RPR	retarding principle rate
ARD	anisotropic rotary diffusion
iARD	improved anisotropic rotary diffusion
pARD	principal anisotropic rotary diffusion
MRD	Moldflow rotary diffusion
NAT	natural closure
FEC	fast exact closure
IBOF	invariant-based optimal fitted
EBOF	eigenvalue-based optimal fitted
ORF	orthotropic fitted
OWE	orthotropic fitted closure approximation for wide interaction coefficients
OWE3	orthotropic fitted closure approximation for wide interaction coefficients with third order polynomial approximation
SPH	smoothed particle hydrodynamic
DEM	discrete element method
pFEA	particle finite element analysis
DNS	direct numerical simulation
MPS	moving particle semi-implicit
EBG	element bending group

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