Measures describing curvilinear short fiber distributions

Received: 9 February 2020 / Accepted: 24 August 2020 / Published online: 8 September 2020
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Abstract In this article, we discuss measures for fibers having a curvilinear shape. This is the case, for example, for man-made cellulose fibers having a weak stiffness. The fibers are bent during the injection molding process of short fiber reinforced plastics. For this purpose, μ-CT data can be evaluated and several measures can be introduced defining the geometrical orientation of the fibers. These measures are the length, a mean curvature, and the mean torsion. Furthermore, a mean orientation of a fiber and a mean deviation to a straight line can be defined. Additionally, to these measures, which are based on a continuous interpolation of given data points, discretized quantities only considering the data points are compared. Finally, the distributions of these measures at real μ-CT data are provided.

Keywords Short fibers · Composite · Fiber orientation · CT analysis · Natural fibers

1 Introduction

Natural fibers are of particular interest in view of sustainability developments, see [6]. Man-made cellulose fiber reinforced injection molded components are an interesting alternative to glass fiber reinforced plastics for certain applications due to their high mechanical properties (elongation at break: 13%) and lower density (1.5 g cm$^{-3}$). They also have a lower bending stiffness than glass fibers. Man-made cellulose fibers are bent during the injection molding process which leads to faster fiber entanglement and a lower fiber preferred orientation, see [9]. In other words, the lower bending stiffness does also not imply alignment as a straight fiber resulting in an anisotropy of the component properties. Furthermore, it is known, see, for example [14], that a fiber reinforced specimen has different material properties over its thickness, which is caused—apart from the melt flow behavior of the thermoplastic—by the spatial distribution of the fiber orientation. Applications from glass fiber reinforced plastic, where the short fibers are straight lines, are discussed in a number of publications, see, for example [7,8,13,18]. In these publications, the fiber orientation of glass fiber reinforced samples has already been extensively investigated. There, the fiber orientation and the fiber length are mainly evaluated. However, a curved fiber has no unique orientation. Furthermore, it can reach various geometrical forms. Thus, measures, principally describing the specific form, are of particular interest.

Since μ-CTs can provide coordinates of each fiber, these can be drawn on to apply interpolation concepts so that a continuous function is assigned to each fiber. These functions can be evaluated, i.e. geometrical measures...
can be provided. There are two possibilities: first, the simplest approach is to draw on polygon approximations using the coordinates describing the fiber, which is easy to be implemented. Second, continuous formulations can be applied implying further measures such as curvature, and torsion. The latter terms will be explained in detail in Sect. 2. We draw on a spline concept to interpolate the coordinates so that each fiber obtains a continuous representation, and, accordingly, the possibility to apply concepts in differential geometry.

The article is structured as follows: since we are interested to introduce measures of curvilinear fibers, a summary of the description of the local behavior of curves is provided. In a second step, global mean values are proposed. These mean values are the orientation, curvature, torsion, and a mean deviation from a straight line. Afterwards, one special interpolation method, which is based on classical, natural splines, is drawn on to evaluate the proposed measures. Afterwards, a discretized, simplified version is offered. Both schemes are compared using real μ-CT data of a man-made cellulose fiber reinforced polypropylene specimen. In this article, the distributions of the measures are discussed.

2 Measures of curvilinear curves

First, we recap some basics in the description of curves, and, subsequently, some measures of curvilinear fibers are proposed characterizing their orientation, curvature and torsion.

2.1 Properties of curves

A curve in the three-dimensional case is described by the position vector \( \mathbf{r}(\xi) \), \( \mathbf{r} \in \mathbb{R}^3 \), which depends on a parameter \( \xi \), \( \xi \in [\xi_1, \xi_2] \). The tangent vector is given by

\[
\mathbf{g}_\xi(\xi) = \frac{d\mathbf{r}}{d\xi} = \mathbf{r}_\xi, \quad \mathbf{g}_\xi \in \mathbb{R}^3.
\]

see, for example, \([3, 11]\). The arc length

\[
s(\xi) = \int_{\xi_1}^{\xi} \| \mathbf{g}_\xi(\xi) \| \, d\xi, \quad s \in [0, L]
\]

represents the length of the covered path from point \( \xi_1 \) until point \( \xi \) on the curve, whereas

\[
L = \int_{\xi_1}^{\xi_2} \| \mathbf{g}_\xi(\xi) \| \, d\xi,
\]

denotes the length of the curve between two points \( \xi_1 \) and \( \xi_2 \) (\( \xi_2 > \xi_1 \)), see Fig. 1a.

\[
s_{s, \xi} = \| \mathbf{g}_\xi \| = \sqrt{\mathbf{g}_\xi \cdot \mathbf{g}_\xi} \neq 0
\]
defines the rate of the arc length representing the norm of the tangent vector or the square root of the metric coefficient concerned. Since \( s, \xi \neq 0 \) holds, there exists a function \( \xi(s) \) having the derivative

\[
\frac{d\xi}{ds} = \frac{1}{\|g_\xi\|}. \tag{5}
\]

In this sense, we will need the mapping between the parameter \( \xi \) and the arc length \( s \). In this case, the curve can also be related to the arc length

\[
\hat{r}(s(\xi)) = r(\xi), \tag{6}
\]

so that the tangent vector (1) reads

\[
ge_\xi = r_\xi s, \xi = \hat{r}_s \|g_\xi\| \tag{7}
\]

leading to the obvious property

\[
e_t(s) := \frac{g_\xi}{\|g_\xi\|} = \frac{r_\xi}{\|r_\xi\|} = \hat{r}_s, \quad \text{i.e. } \|e_t(s)\| = 1. \tag{8}
\]

\( e_t \) is called the unit tangent vector. Using the property

\[
\frac{d}{ds}(e_t \cdot e_t) = 2e_t \cdot e_t_s = 0, \tag{9}
\]

we can see that the vector \( e_{t,s} = r_{ss} \) is orthogonal to \( e_t \), see Eq. (8). This property is chosen to define the unit normal vector

\[
e_n := \frac{e_{t,s}}{\|e_{t,s}\|} = \frac{\hat{r}_{ss}}{\|\hat{r}_{ss}\|}. \tag{10}
\]

The norm of \( \hat{r}_{ss} \) represents the curvature of the curve

\[
\kappa(s) := \|\hat{r}_{ss}\| = \|e_{t,s}\|. \tag{11}
\]

The two vectors \( e_t \) and \( e_n \) are chosen to define the binormal unit vector

\[
e_b(s) := e_t(s) \times e_n(s), \tag{12}
\]

see Fig. 1a, i.e. the natural basis or trihedron of the curve. Using the Frenet equations, see [11],

\[
e_{t,s} = \kappa e_n, \quad e_{n,s} = -\kappa e_n + \tau e_b, \quad e_{b,s} = -\tau e_n \tag{13}
\]

motivates an additional local measure—apart the curvature—namely the torsion \( \tau \). Eq. (13)_1 is already provided by Eq. (10) using definition (11), i.e. \( \kappa \) represents the (local) deviation of the curve from a straight line. \( \tau \) is a measure for the (local) deviation of a curve from a flat course, which can be seen at Eq. (13)_2,3, where the curve remains in the \((e_t, e_n)\)-plane. The torsion itself can be calculated by

\[
\tau = e_{n,s} \cdot e_b = -e_{b,s} \cdot e_n, \tag{14}
\]

see Eq. (13)_2,3. Its sign represents the rotation (right- or left handed).

All these expressions look quite simple if an arc length representation is given. However, later on only the parametric representation \( r(\xi) \) is provided, which requires a different representation. Here, we follow the presentation in [2]. For the subsequent calculations,

\[
\frac{d}{d\xi}\|r_\xi\|^n = n\|r_\xi\|^{n-2}(r_\xi \cdot r_{\xi\xi}) \tag{15}
\]

is required. Then, the product and chain rule leads to

\[
e_{t,s} = \hat{r}_{ss} = \frac{d}{ds}(r_\xi \|r_\xi\|^{-1}) = \frac{1}{\|r_\xi\|}(r_\xi r_{\xi\xi} - (r_\xi \cdot r_{\xi\xi}) r_\xi), \tag{16}
\]

see Eqs. (5) and (7). This expression is necessary to determine the curvature in Eq. (13)_1 by calculating the cross product with \( e_t \)

\[
e_{t,s} \times e_t = \kappa e_n \times e_t = -\kappa e_b, \tag{17}
\]
i.e. using the norm of both sides, we obtain after some calculations

\[ \kappa = \| e_{t,s} \times e_t \| = \frac{1}{\| r_{\xi} \|} \| r_{\xi} \times r_{\xi\xi} \|. \]  

(18)

Now, using again Eq. (13)1 with (16) and (18), we arrive at the representation for the unit normal vector

\[ e_n = \frac{\| r_{\xi} \|}{\| r_{\xi} \times r_{\xi\xi} \|} r_{\xi\xi} - \frac{1}{\kappa} \frac{e_t}{\| e_t \times e_{t,s} \|}. \]  

(19)

With the help of Eqs. (8) and (19), the biaxial unit vector (12) reads

\[ e_b = \frac{r_{\xi} \times r_{\xi\xi}}{\| r_{\xi} \times r_{\xi\xi} \|}. \]  

(20)

Finally, we would like to express the torsion \( \tau \) by the parameter representation. For this purpose, Eq. (13)3 is chosen and the binormal unit vector (12) is inserted

\[ \tau e_n = -e_b, \text{ s.e. } \frac{d}{ds} e_t = \kappa e_t \times e_{t,s} - \frac{1}{\kappa} e_t \times e_{t,ss}. \]  

(21)

This is multiplied with \( e_n = (1/\kappa)e_{t,s} \) leading to

\[ \tau = \frac{1}{\kappa^2} (e_t \times e_{t,s}) \cdot e_{t,ss}. \]  

(22)

i.e. with Eq. (16) and the chain-rule we arrive at an expression of the form

\[ e_{t,ss} = \frac{d}{ds} e_{t,s} = \frac{d}{d\xi} e_{t,s} = \alpha r_{\xi} + \beta r_{\xi\xi} + \gamma r_{\xi\xi\xi}. \]  

(23)

(Here, we need only the term \( \gamma = \| r_{\xi} \|^{-3} \), whereas \( \alpha \) and \( \beta \) will vanish in the subsequent calculations.) We insert Eqs. (16) and (23) into Eq. (22) yielding

\[ \tau = \frac{1}{\kappa^2} \| r_{\xi} \| (r_{\xi} \times r_{\xi\xi}) \cdot r_{\xi\xi\xi} = \frac{(r_{\xi} \times r_{\xi\xi}) \cdot r_{\xi\xi\xi}}{\| r_{\xi} \times r_{\xi\xi} \|^2}. \]  

(24)

with \( \kappa \) from Eq. (18). Thus, all required quantities are provided in dependence of the parameter \( \xi \).

2.2 Measures defining a straight line (fiber)

Short glass or carbon fibers can be modeled as straight lines. For a straight line, \( \hat{r}(s) = r_1 + se_t \), we have \( \kappa = 0 \), and \( \tau \) is defined to be zero (\( e_n \) is in this case arbitrary, but orthonormal to \( e_t \)), i.e. only the length \( L \) and the orientation \( e_t \) are chosen for characterizing purposes. The orientation is given by Eq. (8), \( r_{t,s} = e_t \), see Fig. 2. Thus, we need the starting and the end point of a fiber, \( r_1 \) and \( r_2 \), respectively,

\[ e_t = \frac{r_2 - r_1}{\| r_2 - r_1 \|}, \quad L = \| r_2 - r_1 \|. \]  

(25)

The orientation vector \( e_t \) can be expressed in a spherical coordinate system by the two angles \( \varphi \) and \( \vartheta \),

\[ e_t = \sin \vartheta \cos \varphi e_1 + \sin \vartheta \sin \varphi e_2 + \cos \vartheta e_3, \]  

(26)

see, for example, [12]. Frequently, the two angles

\[ \vartheta = \arccos |e_t \cdot e_3|, \quad \varphi = \arccos \left| \frac{e_t \cdot e_1}{\sin \vartheta} \right| \]  

(27)

are chosen for characterizing the orientation of fibers in composites, see, for example, [1, 17]. Thus, straight fibers are simple to be evaluated.
2.3 Measures defining a curve

Since we are interested in scalar measures representing some “mean-value” of measures characterizing curvilinear fibers, we introduce two approaches. First, given data points representing a spatial curve are interpolated to obtain a continuous representation (piece-wise $C^\infty$, but globally $C^2$ continuous). Based on these approximations, the measures (mean curvature, torsion, direction, and deviation from straight line) are calculated. Afterwards, these measures are provided without the interpolation concept, i.e. only the data points are evaluated directly.

2.3.1 Mean values using continuous interpolation

Before defining the mean values of interest, a fiber—given by $n_d$ data points $d_k, k = 1, \ldots, n_d$—should be represented by an interpolating function,

$$r(\xi) = \hat{x}_k(\xi)e_k = \hat{x}_1(\xi)e_1 + \hat{x}_2(\xi)e_2 + \hat{x}_3(\xi)e_3.$$  \hspace{1cm} (28)

As a first approach, piece-wise polynomials of the third order are chosen,

$$\hat{x}_{ki}(\xi) = a_{ki} + b_{ki}(\xi - \xi_i) + c_{ki}(\xi - \xi_i)^2 + d_{ki}(\xi - \xi_i)^3, \quad \text{for} \quad \xi_i \leq \xi < \xi_{i+1}, \quad i = 1, \ldots, n_d - 1,$$  \hspace{1cm} (29)

where the sampling points $\xi_i$ are estimated using the Chord-approach

$$\xi_i = \sum_{j=1}^{i-1} \|d_{j+1} - d_j\|, \quad i = 2, \ldots, n_d,$$  \hspace{1cm} (30)

with $\xi_1 = 0$. The index $i$ defines the interval, $i = 1, \ldots, n_d - 1$, and $k = 1, 2, 3$ the coordinate direction. In other words, we have $\hat{x}_k(\xi) = \hat{x}_{ki}(\xi)$ in interval $i, \xi_i \leq \xi < \xi_{i+1}$. Obviously, we obtain for $\xi = \xi_i$ the parameters

$$a_{ki} = d_i \cdot e_k = d_{ik}, \quad i = 1, \ldots, n_d - 1$$  \hspace{1cm} (31)

of each coordinate, $k = 1, 2, 3$ (starting point implies $r(\xi_1) = d_1$). Using this approach, the first three derivatives can be determined

$$\hat{x}_k''(\xi) = b_{ki} + 2c_{ki}(\xi - \xi_i) + 3d_{ki}(\xi - \xi_i)^2,$$

$$\hat{x}_k'''(\xi) = 2c_{ki} + 6d_{ki}(\xi - \xi_i),$$

$$\hat{x}_k''''(\xi) = 6d_{ki}. \hspace{1cm} (32)$$
In each interval $i$, four unknowns are given, i.e. we have $4 \times (n_d - 1)$ unknowns. Thus, we need the same number of equations. With Eq. (31) and the end point $r(\xi_{n_d}) = d_{n_d}$, there are $n_d$ equations. Furthermore, we require that the following conditions are fulfilled at the point $\xi_{i+1}$

$$\dot{x}_k(\xi_{i+1}) = \dot{x}_{ki+1}(\xi_{i+1}), \quad \ddot{x}_k(\xi_{i+1}) = \ddot{x}_{ki+1}(\xi_{i+1}), \quad \dot{x}_k''(\xi_{i+1}) = \dot{x}_{ki+1}''(\xi_{i+1}), \quad (33)$$

$i = 1, \ldots, n_d - 2$, leading to $3 \times (n_d - 2)$ equations. Thus, two additional equations are required. Here, we draw on natural splines, i.e. we assume curvature free starting and ending points

$$\ddot{x}_k''(\xi_1) = 0, \quad \ddot{x}_{kn_d-1}(\xi_{n_d}) = 0. \quad (34)$$

The index $ki + 1$ in Eq. (33) defines the $k$th component (or its derivatives), see Eq. (28), in interval $i + 1$. To avoid a comma separation, which is chosen for derivatives, a comma-free notation is chosen. This holds for $kn_d - 1$ in Eq. (34) as well, i.e. the evaluation of the $k$th function defined in interval $n_d - 1$. The numerical implementation to solve the linear system of equations is provided in [5], see [4] as well. In other words, the given data points are exactly fulfilled. The curves are continuous of second order. In conclusion, the parameters $a_{ki}, b_{ki}, c_{ki}, d_{ki}, i = 1, n_d - 1, k = 1, 2, 3$, are assumed to be known in the following. This implies that in each interval the position vector and its derivatives

$$\mathbf{r}_i(\xi) = \dot{x}_k(\xi)\mathbf{e}_k, \quad \mathbf{r}_{i,\xi}(\xi) = \ddot{x}_k(\xi)\mathbf{e}_k, \quad \mathbf{r}_{i,\xi\xi}(\xi) = \dddot{x}_k(\xi)\mathbf{e}_k, \quad \mathbf{r}_{i,\xi\xi\xi}(\xi) = \ddddot{x}_k(\xi)\mathbf{e}_k \quad (35)$$

are known in each interval $i, \xi_i \leq \xi < \xi_{i+1}, i = 1, \ldots, n_d - 1$.

In the following, we define the measures determining fibers. First, the length of a fiber has to be computed using definition (3)

$$L = \int_0^{\xi_{n_d}} \| \mathbf{r}_{i,\xi}(\xi) \| \, d\xi = \sum_{i=1}^{n_d-1} \int_{\xi_i}^{\xi_{i+1}} \| \mathbf{r}_{i,\xi}(\xi) \| \, d\xi. \quad (36)$$

The integrals can, for example, be evaluated using Simpson’s rule, see, for instance, [16], with $\Delta \xi = (\xi_{i+1} - \xi_i)/n_S, n_S$ is the number of subintervals between two data points. Then, we have to calculated at the integration points $\xi_i(\ell) = \xi_i + (\ell - 1)\Delta \xi$,

$$\int_{\xi_i}^{\xi_{i+1}} f(\xi) \, d\xi \approx \frac{\Delta \xi}{6} \sum_{\ell=1}^{n_S} \left( f(\xi_i^{(\ell)}) + 4f(\xi_i^{(\ell)}) + f(\xi_i^{(\ell)} + \Delta \xi) \right). \quad (37)$$

In the following, all integrals are evaluated using this integration scheme. (Of course, other schemes or adaptive quadrature rules are possible.)

Next, the mean curvature

$$\kappa_M := \frac{\int_0^{\xi_{n_d}} \kappa(\xi)\|\mathbf{r}_{i,\xi}(\xi)\| \, d\xi}{\int_0^{\xi_{n_d}} \|\mathbf{r}_{i,\xi}(\xi)\| \, d\xi} = \frac{1}{L} \int_0^{\xi_{n_d}} \frac{\|\mathbf{r}_{i,\xi} \times \mathbf{r}_{i,\xi\xi}\|}{\|\mathbf{r}_{i,\xi}\|^2} \, d\xi, \quad (38)$$

the mean torsion

$$\tau_M := \frac{1}{L} \int_0^{\xi_{n_d}} \frac{\mathbf{r}_{i,\xi} \times \mathbf{r}_{i,\xi\xi} \cdot \mathbf{r}_{i,\xi\xi\xi}}{\|\mathbf{r}_{i,\xi} \times \mathbf{r}_{i,\xi\xi}\|^2} \|\mathbf{r}_{i,\xi}(\xi)\| \, d\xi, \quad (39)$$

and the mean orientation (direction)

$$\mathbf{r}_{M,\xi} := \frac{1}{L} \int_0^{\xi_{n_d}} \mathbf{r}_{i,\xi}(\xi) \|\mathbf{r}_{i,\xi}(\xi)\| \, d\xi \quad (40)$$

are defined, see Eqs. (18) and (24). In this sense, the two angles (27) can be chosen for a mean orientation,

$$\vartheta_M = \arccos \left| \frac{\mathbf{r}_{M,\xi} \cdot \mathbf{e}_3}{\|\mathbf{r}_{M,\xi}\|} \right|, \quad \varphi_M = \arccos \left| \frac{\mathbf{r}_{M,\xi} \cdot \mathbf{e}_1}{\sin \vartheta_M} \right|. \quad (41)$$

An additional quantity is the deviation of the curved line to a straight line. The straight line is defined by

$$\mathbf{r}_L(\lambda) = \mathbf{r}_C + \lambda \mathbf{r}_{M,\xi}, \quad (42)$$
Fig. 3 Center point of a line and the straight line approximation of a curve

where

$$r_C := \frac{1}{L} \int_0^{\xi_n} \|r(\xi)\| \, d\xi$$

(43)
defines the center of the line (fiber), see Fig. 3. The distance of a point of the line \(r(\xi)\) to the straight line \(42\) is given by

$$d(\xi) = \left\| (r_C - r(\xi)) \times \frac{r_{M,\xi}}{\|r_{M,\xi}\|} \right\|,$$

(44)

see, for instance, [10], so that the mean value is defined by

$$d_M := \frac{1}{L} \int_0^{\xi_n} d(\xi) \left\| \frac{r_C - r(\xi)}{\|r_{M,\xi}\|} \right\| \, d\xi = \frac{1}{L} \int_0^{\xi_n} \left\| (r_C - r(\xi)) \times \frac{r_{M,\xi}}{\|r_{M,\xi}\|} \right\| \, d\xi.$$  

(45)

In conclusion, we have several measures defining a curve (length \(L\) (36), mean curvature \(\kappa_M\) (38), mean torsion \(\tau_M\) (39), mean orientations \((\vartheta_M, \varphi_M)\) (41), and the mean deviation \(d_M\) (45) from the mean orientation line (42)).

Particularly, the length might be of interest to estimate—together with a mean diameter—the volume of the fiber, or the orientation vector \(n = r_{M,\xi} / \|r_{M,\xi}\|\) in Eq. (40) is drawn on to determine an orientation tensor \(N = n \otimes n\), see [15]. However, new measures such as mean curvature or mean deviation to a straight line give insight into the flow behavior and the weakness of natural fibers relative to the matrix material.

2.3.2 Mean values using simplified approach

Apart from the more precise definitions using interpolated points in Sect. 2.3.1, an alternative, simplified access drawing on rough approximations of the derivatives by numerical differentiation and the integrals applying a trapezoidal rule can be provided.

Again the chord method (30) is applied using

$$\Delta \xi_i := \|d_{i+1} - d_i\|, \quad i = 1, \ldots, n_d - 1.$$  

(46)

In a first step, we compute at all data points \(i = 1, \ldots, n_d\) the first three derivatives using the approximations in “Appendix A”. The tangent vector reads, according to Eq. (65),

$$r_{i,\xi} (\xi_i) \approx r_{i,\xi} := \frac{\Delta \xi_i^2}{D_i} \Delta d_i + \frac{\Delta \xi_i^2}{D_i} \Delta d_{i-1}, \quad i = 2, \ldots, n_d - 1$$

(47)

with

$$\Delta d_i = d_{i+1} - d_i, \quad D_i = \Delta \xi_{i-1} \Delta \xi_i (\Delta \xi_{i-1} + \Delta \xi_i), \quad \Delta \xi_i = \xi_{i+1} - \xi_i.$$  

(48)

The tangent vectors at the first and the last point can be estimated using Eq. (67),

$$r_{1,\xi} = \frac{\Delta d_1}{\Delta \xi_1}, \quad r_{n_d,\xi} = \frac{\Delta d_{n_d-1}}{\Delta \xi_{n_d-1}}.$$  

(49)

The second derivative reads according to Eq. (70)

$$r_{i,\xi \xi} = A_i (\Delta d_i - \Delta d_{i-1}) + B_i r_{i,\xi}, \quad i = 2, \ldots, n_d - 1,$$

(50)
Table 1  Approximated measures characterizing a curvilinear fiber with the differential formulas (47), (49), (50), and (52)–(54)

| Purpose                              | Formula                                                                 |
|--------------------------------------|-------------------------------------------------------------------------|
| Length                               | \( L = \frac{1}{2} \sum_{i=1}^{n_d-1} \left( \| r_{i,\xi} \| + \| r_{i+1,\xi} \| \right) \Delta \xi_i \) (57)         |
| Mean curvature \( \kappa_M \)        | \( \kappa_M = \frac{1}{2L} \sum_{i=1}^{n_d-1} \left( \frac{\| r_{i,\xi} \times r_{i,\xi+1} \|}{\| r_{i,\xi} \| \| r_{i+1,\xi+1} \|} + \frac{\| r_{i+1,\xi+1} \times r_{i,\xi+1} \|}{\| r_{i,\xi} \| \| r_{i+1,\xi+1} \|} \right) \Delta \xi_i \) (58) |
| Mean torsion \( \tau_M \)           | \( \tau_M = \frac{1}{2L} \sum_{i=1}^{n_d-1} \left( \frac{\| r_{i,\xi} \times r_{i,\xi+1} \|}{\| r_{i,\xi} \| \| r_{i+1,\xi+1} \|} \| r_{i,\xi} \| + \frac{\| r_{i+1,\xi} \times r_{i,\xi+1} \|}{\| r_{i,\xi} \| \| r_{i+1,\xi+1} \|} \| r_{i+1,\xi+1} \| \right) \Delta \xi_i \) (59) |
| Mean direction \( r_{M,\xi} \)      | \( r_{M,\xi} = \frac{1}{2L} \sum_{i=1}^{n_d-1} \left( r_{i,\xi} \| r_{i,\xi} \| + r_{i+1,\xi} \| r_{i+1,\xi} \| \right) \Delta \xi_i \) (60) |
| Center                               | \( r_C = \frac{1}{2L} \sum_{i=1}^{n_d-1} \left( d_i \| r_{i,\xi} \| + d_{i+1} \| r_{i+1,\xi} \| \right) \Delta \xi_i \) (61) |
| Mean deviation from a straight line  | \( d_M = \frac{1}{2L} \sum_{i=1}^{n_d-1} \left( \| r_C - d_i \| \left( \| r_{i,\xi} \| + \| r_{C - d_{i+1}} \times \frac{r_{M,\xi}}{\| r_{M,\xi} \|} \right) \right) \Delta \xi_i \) (62) |

with the abbreviations

\[
A_i = \frac{2}{\Delta \xi_{i-1}^2 + \Delta \xi_i^2}, \quad B_i = \Delta \xi_{i-1} - \Delta \xi_i.
\] (51)

Here, too, we must estimate the derivatives at the first and last point, see Eq. (71),

\[
r_{1,\xi} = \frac{r_{2,\xi} - r_{1,\xi}}{\Delta \xi_1}, \quad r_{n_d,\xi} = \frac{r_{n_d,\xi} - r_{n_d-1,\xi}}{\Delta \xi_{n_d-1}}.
\] (52)

Finally, the third derivative of the spatial curve is estimated using Eq. (72)

\[
r_{i,\xi} = \frac{\Delta \xi_i^2}{D_i} (r_{i+1,\xi} - r_{i,\xi}) + \frac{\Delta \xi_i^2}{D_i} (r_{i,\xi} - r_{i-1,\xi}), \quad i = 2, \ldots, n_d - 1,
\] (53)

where the third derivatives at both the first as well as last point have to be estimated, see Eq. (73),

\[
r_{1,\xi} := \frac{r_{2,\xi} - r_{1,\xi}}{\Delta \xi_1}, \quad r_{n_d,\xi} := \frac{r_{n_d,\xi} - r_{n_d-1,\xi}}{\Delta \xi_{n_d-1}}.
\] (54)

Obviously, numerical differentiation makes higher-order derivatives more and more inaccurate.

Since all derivatives are known at the data points, the measures of Sect. 2.3.1 can be calculated. For this purpose, a simple numerical integration scheme has to be applied. Since the trapezoidal rule

\[
\int_{\xi=0}^{\xi_{n_d}} f(\xi) \, d\xi \approx \sum_{i=1}^{n_d-1} \frac{f(\xi_i) + f(\xi_{i+1})}{2} \Delta \xi_i = \frac{1}{2} (f(\xi_1) \Delta \xi_1 + f(\xi_{n_d}) \Delta \xi_{n_d-1}) + \sum_{i=2}^{n_d-1} f(\xi_i) \Delta \xi_i,
\] (55)

is more accurate—for nearly the same computational work—than the left rectangular integration scheme

\[
\int_{\xi=0}^{\xi_{n_d}} f(\xi) \, d\xi \approx \sum_{i=1}^{n_d-1} f(\xi_i) \Delta \xi_i,
\] (56)

use is made of the trapezoidal rule to compute the integrals.

The mean measures using the trapezoidal rule are assembled in Table 1.
3 Example

In the following, we apply the two concepts to real μ-CT data of a man-made cellulose fiber reinforced polypropylene specimen. The data are generated using a 3D X-ray microscope Xradia 520 Versa from Zeiss. The fibers were detected with the program Avizo (version 9.4.0 of the company FEI). The considered volume element of the sample has a size of approximately 208.5 μm × 629.6 μm × 629.6 μm. This produces the fibers shown in Fig. 4. Here, we are not interested in discussing whether it makes sense considering fibers which are cut at the edge surfaces of the cube-like region or not. We are interested in how the applied schemes work with some data.

To obtain more insight in the fiber distribution, we draw on the concepts of Sects. 2.3.1 and 2.3.2. Figures 5 and 6 compile the results. The length computation shows very similar results of both methods, the interpolation concept, and the simplified scheme, see Fig. 5a and b.

The mean curvature in Fig. 5c and d, however, shows slightly different results, since the simplified method draws on the second derivative, and derivatives using differential formulas roughen the results. This is the essential drawback in the mean torsion computation shown in Fig. 5e–f, where the third derivative of the curve vector is evaluated. Since there are some fibers with “kinks”, totally unrealistic values are computed using the simplified version to determine the torsion of a fiber. The estimation of the mean fiber orientation, represented by the angles ϑ_M and ϕ_M, is appropriate for both schemes, see Fig. 6a–d, which holds for the mean distance of the fibers relative to the center line (42) in Fig. 6e–f.

4 Conclusions

In this paper, the quantitative properties of man-made cellulose fibers are proposed. Since these fibers represent curvilinear functions in the matrix material, some characteristic measures are of particular interest. We propose several measures describing curvilinear lines: length, mean curvature, mean torsion, mean direction, and a mean distance to a straight line. To describe the curvilinear fibers, the discrete data points are interpolated by a first concept. This interpolation is based on splines to obtain a continuous representation. In a second scheme, only the discrete data points are evaluated. There, the required first three derivatives are based on differential derivative formulas leading to a much simpler evaluation process. However, it turns out that the third derivative leads to very inadequate results in some situations so that the measure of mean torsion, i.e., a measure of the deviation of a curve from a plane, is not suitable. All other measures lead—for the distribution of a real μ-CT scan data—to similar results. Thus, both methods are a promising tool for evaluating the fiber measures, except for torsion, where only the more sophisticated evaluation procedure is necessary.
Fig. 5 Distribution of fiber length, mean curvature, and torsion

(a) Length distribution (spline approach)

(b) Length distribution (simplified)

(c) Mean curvature distribution (spline approach)

(d) Mean curvature distribution (simplified)

(e) Mean torsion distribution (spline approach)

(f) Mean torsion distribution (simplified)
Fig. 6 Distribution of mean orientation angles and mean distance to straight line

Acknowledgements We would like to thank Dr. Matthias Grafenhorst for some discussion in programming, and to Prof. Dr.-Ing. A. Ries for her support.

Funding Open Access funding provided by Projekt DEAL.

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Thus, it has to be explained how the derivatives are approximated. Here, several possibilities exist, see, for example, [4, 19], most commonly derived for equidistant point spacing.

A Approximation of derivatives by differential formulas

Concerning the approximation of the vectors \( \mathbf{r}, \mathbf{\xi}, \mathbf{r} \times \mathbf{\xi}, \) and \( \mathbf{r} \times \mathbf{\xi} \times \mathbf{\xi} \), we draw on concepts of differential formulas and motivate the procedure at a scalar function. First, we look at the approximation of the first derivative. Even here, several possibilities exist, see, for example, [4, 19], most commonly derived for equidistant point spacing. Thus, it has to be explained how the derivatives are approximated.

If we interpolate three points \((\xi_{i-1}, y_{i-1}), (\xi_i, y_i),\) and \((\xi_{i+1}, y_{i+1})\) with a quadratic polynomial,

\[
 f(\xi) = a_0 + a_1(\xi - \xi_i) + a_2(\xi - \xi_i)^2,
\]

the three conditions \( f(\xi_{i-1}) = y_{i-1}, f(\xi_i) = y_i, \) and \( f(\xi_{i+1}) = y_{i+1} \) yield the coefficients

\[
 a_0 = y_i, \\
 a_1 = \frac{\Delta \xi_i^2 (y_{i+1} - y_i) + \Delta \xi_i^2 (y_i - y_{i-1})}{\Delta \xi_{i-1} \Delta \xi_i (\Delta \xi_{i-1} + \Delta \xi_i)} \\
 a_2 = \frac{\Delta \xi_{i-1} (y_{i+1} - y_i) - \Delta \xi_i (y_i - y_{i-1})}{\Delta \xi_{i-1} \Delta \xi_i (\Delta \xi_{i-1} + \Delta \xi_i)}
\]

with \( \Delta \xi_i = \xi_{i+1} - \xi_i \) and \( \Delta \xi_{i-1} = \xi_i - \xi_{i-1} \). The approximated derivative at \( \xi_i \) reads

\[
 f'_i := f'(\xi_i) = a_1 = \frac{\Delta \xi_i^2 (y_{i+1} - y_i) + \Delta \xi_i^2 (y_i - y_{i-1})}{\Delta \xi_{i-1} \Delta \xi_i (\Delta \xi_{i-1} + \Delta \xi_i)} = \frac{y_i - y_{i-1}}{\Delta \xi_{i-1}} + \frac{\Delta \xi_{i-1} (y_{i+1} - y_i) - \Delta \xi_i (y_i - y_{i-1})}{\Delta \xi_i (\Delta \xi_{i-1} + \Delta \xi_i)},
\]

\( i = 2, \ldots, n_d - 1 \). Since the central differential formula cannot provide a derivative at the first and last point, we draw on a forward differential formula at \( \xi_1 \) and a backward differential formula at the last point \( \xi_{n_d} \),

\[
 f'_1 := \frac{y_2 - y_1}{\Delta \xi_1}, \\
 f'_{n_d} := \frac{y_{n_d} - y_{n_d-1}}{\Delta \xi_{n_d-1}}.
\]

The second derivative can be estimated by a Taylor series around \( \xi_i \) to the right and to the left side:

\[
 f(\xi_{i+1}) = f(\xi_i) + \Delta \xi_i f'(\xi_i) + \frac{\Delta \xi_i^2}{2} f''(\xi_i) + \cdots \\
 f(\xi_{i-1}) = f(\xi_i) - \Delta \xi_i f'(\xi_i) + \frac{\Delta \xi_i^2}{2} f''(\xi_i) + \cdots
\]

We take the sum of both equations, neglect the higher-order terms, and rearrange the result to obtain the second derivative:

\[
 f''(\xi_i) \approx 2 \frac{f(\xi_{i-1}) - 2 f(\xi_i) + f(\xi_{i+1})}{\Delta \xi_{i-1}^2 + \Delta \xi_i^2} + (\Delta \xi_{i-1} - \Delta \xi_i) f'(\xi_i),
\]

\( i = 2, \ldots, n_d - 1 \). This can be rewritten and approximated using Eq. (66)

\[
 f''_i = 2 \frac{y_{i-1} - 2 y_i + y_{i+1}}{\Delta \xi_{i-1}^2 + \Delta \xi_i^2} + (\Delta \xi_{i-1} - \Delta \xi_i) f'_i,
\]

\( i = 1, \ldots, n_d - 1 \). If \( \Delta \xi_{i-1} \approx \Delta \xi_i \), the last term might be neglected. However, this cannot be guaranteed in the application here. To estimate the second derivative at the first and the last point, again forward and backward differential formulas (linear interpolation) are applied,

\[
 f''_1 := \frac{f'_2 - f'_1}{\Delta \xi_1}, \\
 f''_{n_d} := \frac{f'_{n_d} - f'_{n_d-1}}{\Delta \xi_{n_d-1}}.
\]
The third derivative can be computed by inserting the second derivative into Eq. (65) instead of the function values (quadratic interpolation)

\[
f''_i := \frac{\Delta \xi_{i-1}^2 (f''_{i+1} - f''_i) + \Delta \xi_i^2 (f''_i - f''_{i-1})}{\Delta \xi_{i-1} \Delta \xi_i (\Delta \xi_{i-1} + \Delta \xi_i)}
\]

(72)

i = 3, \ldots, n_d - 2; thus, five points are required at least to estimate the third derivative (the values f''_{i+1} requires \( y_{i+2} \) and f''_{i-1} requires \( y_{i-2} \) if the differentials (70) would be used. Since we have estimated the second derivatives at the first and last point, see Eq. (71), formula (72) can be evaluated also in the range i = 2, \ldots, n_d - 1. The third-order derivatives at the “starting” and “ending” points are estimated by (linear interpolation)

\[
f''_1 := \frac{f'_2 - f'_1}{\Delta \xi_1}, \quad f''_{n_d} := \frac{f''_{n_d-1} - f''_n}{\Delta \xi_{n_d-1}}.
\]

(73)

Therefore, at each point \( \xi_i, \, i = 1, \ldots, n_d \), a derivative can be estimated.

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