ON THE DILATED FACETS OF A POISSON-VORONOI TESSELLATION

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ABSTRACT

In this paper, the parallel set $\xi_R$ of the facets ($(d-1)$-faces) of a stationary Poisson-Voronoi tessellation in $\mathbb{R}^2$ and $\mathbb{R}^3$ is investigated. An analytical formula for the spherical contact distribution function of the tessellation allows for the derivation of formulae for the volume density and the specific surface area of $\xi_R$. The densities of the remaining intrinsic volumes are studied by simulation. The results are used for fitting a dilated Poisson-Voronoi tessellation to the microstructure of a closed-cell foam.

Keywords: contact distribution function, intrinsic volumes, parallel set, surface density, stochastic geometry.

INTRODUCTION

Random tessellations are widely used to model cellular or polycrystalline materials such as foams or sintered ceramics (see, e.g., Coster et al. (2005); Redenbach (2009); Ribeiro-Ayeh (2005); Telley et al. (1996)). For instance, the system of facets of a random tessellation can model the wall system in closed-cell foams or the grain boundaries in polycrystalline materials (see Fig. 1 for examples of such materials). Random tessellations are space-filling cell systems by definition, i.e., their cell facets have zero thickness. In contrast, real materials typically consist of two components such that the facets have a certain thickness. Models for such microstructures can be obtained using the parallel set (or dilation) of the facets of the tessellation’s cells. Using these models, relations between the microstructure of a material and its macroscopic properties can be investigated (Redenbach, 2009).

Fig. 1. Sectional microscopic image of an Al$_2$O$_3$ sinter material (left) and visualisation of a tomographic image of a closed polymer foam (right).

A well-known tessellation model is the Voronoi tessellation generated by a stationary Poisson point process. A huge number of analytic results for geometric characteristics of this model is available (see Okabe et al. (2000) for an overview). Here, we will use explicit formulae for the spherical contact distribution function to derive formulae for the volume and surface area density of the dilated system of cell facets. These characteristics, which are easy to estimate and to interpret, are sufficient for the estimation of the model parameters, the intensity $\lambda$ of the Poisson process and the dilation radius $R$. The densities of the remaining intrinsic volumes can then be used to validate a fitted model. So far, however, they can only be studied by simulation.

The paper is organised as follows: First, we give an introduction into the concepts and the notation used in this paper. Then we obtain some results for the contact distribution function of the random closed set $\xi_R$ formed by the dilated facet system of a stationary Poisson-Voronoi tessellation. In the following section, the densities of the intrinsic volumes of $\xi_R$ are given as functions of the dilation radius $R$. Section “Application” is concerned with the application of the obtained results to the modelling of cellular materials. We conclude with a discussion of the results and an outlook on possible future work.

DILATED POISSON-VORONOI TESSELLATIONS

Throughout this paper we are working in $d$-dimensional Euclidean space $\mathbb{R}^d$ equipped with the Euclidean norm $\| \cdot \|$. For $x \in \mathbb{R}^d$ and $r \geq 0$ let $b(x,r)$ denote the closed $d$-dimensional ball of radius $r$ centred in $x$. The unit sphere in $\mathbb{R}^d$ is denoted by
\(S^{d-1}\). Write \(\mathcal{B}\) for the Borel sets in \(\mathbb{R}^d\), \(\lambda_d\) for the \(d\)-dimensional Lebesgue measure, and \(\sigma_{d-1}\) for the surface measure on \(S^{d-1}\).

Let \(\mathcal{T}\) and \(\mathcal{X}\) be the system of closed and compact convex sets in \(\mathbb{R}^d\), respectively. Elements of \(\mathcal{X}\) are called convex bodies. Equip \(\mathcal{T}\) with the topology of closed convergence (Schneider and Weil, 2008, p. 18) and denote the corresponding system of Borel sets on \(\mathcal{T}\) by \(\mathcal{B}(\mathcal{T})\). A random closed set \(\Xi\) is a measurable mapping from a probability space to \((\mathcal{T}, \mathcal{B}(\mathcal{T}))\). It is called stationary if its distribution is invariant under translations of \(\mathbb{R}^d\).

For sets \(K \in \mathcal{X}\), the intrinsic volumes \(V_k(K)\), \(k = 0, \ldots, d\), can be defined using the Steiner formula

\[
\lambda_d(K + b(0,r)) = \sum_{k=0}^{d} \kappa_k V_{d-k}(K)r^k, \quad r \geq 0.
\]

Here, \(\oplus\) denotes Minkowski addition, i.e., \(A \oplus B = \{a + b : a \in A, b \in B\}\) for \(A, B \in \mathcal{B}\), and \(\kappa_k\) is the volume of the \(k\)-dimensional unit ball. For \(d = 2\) the intrinsic volumes are up to constant factors – the area \(A = V_2\), the boundary length \(L = 2V_1\), and the Euler characteristic \(\chi = V_0\), for \(d = 3\) they are the volume \(V = V_3\), the surface area \(S = 2V_2\), the integral of mean curvature \(M = \pi V_1\), and the Euler characteristic \(\chi = V_0\) (Schneider, 1993, p. 210).

Important characteristics for stationary random closed sets \(\Xi\) are the densities of the intrinsic volumes. They are defined as the limits

\[
V_k(\Xi) = \lim_{r \to \infty} \frac{E V_k(\Xi \cap rW)}{V_d(rW)}, \quad k = 0, \ldots, d, \quad (1)
\]

where \(W \in \mathcal{X}\) is a convex body with \(V_0(W) > 0\) and \(E\) denotes expectation. A sufficient condition for the existence of the limit is that \(\Xi\) is almost surely locally polyconvex, i.e., for \(K \in \mathcal{X}\) the set \(\Xi \cap K\) can almost surely be written as a finite union of convex bodies, and satisfies \(E 2^n(\Xi \cap \mathbb{B}(0,1)^d) < \infty\) (Schneider and Weil, 2008, Theorem 9.2.1). Here \(N(X)\) denotes the smallest number \(m\) such that \(X = K_1 \cup \ldots \cup K_m\) with \(K_1, \ldots, K_m \in \mathcal{X}\).

Analogously to the intrinsic volumes, their densities have the following meaning: For \(d = 2\), \(A_2 = V_{2,2}\) is the area density, \(L_2 = 2V_{1,1}\) is the density of the boundary length, and \(\chi_2 = V_{0,0}\) is the density of the Euler characteristic, for \(d = 3\), \(V_2 = V_{1,3}\) is the volume density, \(S_2 = 2V_{2,1}\) is the surface density (or specific surface area), \(M_V = \pi V_{1,1}\) is the density of the integral of mean curvature, and \(\chi_0 = V_{0,0}\) is the density of the Euler characteristic.

Denote by \(\hat{\mathcal{C}}\) the interior of the bounded set \(C \subset \mathbb{R}^d\). Let \(T\) be a set of bounded convex \(d\)-dimensional subsets of \(\mathbb{R}^d\) with

(i) if \(C_1, C_2 \in T\) and \(C_1 \neq C_2\) then \(\hat{C}_1 \cap \hat{C}_2 = \emptyset\),
(ii) \(\bigcup_{C \in T} C = \mathbb{R}^d\),
(iii) \(T\) is locally finite, i.e., for every bounded \(B \in \mathcal{B}\) it holds that \(\#\{C \in T : C \cap B \neq \emptyset\} < \infty\).

Then \(T\) is called a tessellation of \(\mathbb{R}^d\). The cells are \(d\)-polytopes due to their convexity and the fact that the tessellation is space-filling (Schneider and Weil, 2008, Lemma 10.1.1). Denote by \(\mathcal{B}^k(C)\) the set of all \(k\)-faces, \(k = 0, \ldots, d\), of a \(d\)-polytope \(C\). The space \(\mathcal{T}\) of all tessellations of \(\mathbb{R}^d\) can also be equipped with the topology of closed convergence. Hence, we can define a random tessellation as a random variable \(X\) with values in \(\mathcal{T}\). Then the union of all facets, i.e., \((d-1)\)-faces, of the tessellation forms a random closed set

\[
\Xi = \bigcup_{F \in \mathcal{F}^{d-1}(X)} F,
\]

where \(\mathcal{F}^{d-1}(X) = \bigcup_{C \in X} \mathcal{F}^{d-1}(C)\).

The Poisson-Voronoi tessellation is maybe the most well-known random tessellation model (Schneider and Weil, 2008; Stoyan et al., 1995). It is generated by the points of a stationary Poisson process \(\Phi\) by assigning to a point \(x \in \Phi\) the cell

\[
C(x, \Phi) = \{y \in \mathbb{R}^d : ||y - x|| \leq ||y - z||\} \quad \text{for all } z \in \Phi\}.
\]

In the following, let \(\Phi\) be a stationary Poisson process with intensity \(\lambda\), \(\text{Vo}(\Phi)\) the induced Voronoi tessellation, and \(\Xi\) the stationary random closed set formed by the facets of \(\text{Vo}(\Phi)\). For \(R \geq 0\) denote by

\[
\Xi_R = \Xi \oplus b(0,R) = \bigcup_{F \in \mathcal{F}^{d-1}(\text{Vo}(\Phi))} (F \oplus b(0,R)) \quad (2)
\]

the dilated facet system of \(\text{Vo}(\Phi)\). Finally, denote by \(F_0(0)\) the almost surely unique cell of \(\text{Vo}(\Phi)\) which contains the origin. In the following, we will investigate geometric characteristics of the random closed set \(\Xi_R\). Realisations of this set in both \(\mathbb{R}^2\) and \(\mathbb{R}^3\) are shown in Fig. 2.

![Fig. 2. Realisations of dilated Poisson-Voronoi tessellations in \(\mathbb{R}^2\) (left) and \(\mathbb{R}^3\) (right).](image-url)
CONTACT DISTRIBUTION FUNCTIONS

Let $B \in \mathcal{X}$ be a convex body containing the origin and let $\Theta$ be a random closed set. Then the contact distribution function $H_B$ of $\Theta$ is defined via

$$H_B(r) = P\left(\Theta \cap rB \neq \emptyset | 0 \notin \Theta\right), \quad r \geq 0.$$  

Important special cases are the spherical contact distribution function $H_s$, where $B = b(0,1)$ is the unit ball centred in the origin, and the linear contact distribution function $H_l$, where $B = l(u)$ is a line segment of unit length in direction $u \in S^{d-1}$.

Explicit formulae for contact and chord length distribution functions of the Poisson-Voronoi tessellation are given in Heinrich (1998) and Mache and Stoyan (1992).

**Theorem 1.** (Heinrich, 1998, Theorem 1) Let $B \subset \mathbb{R}^d$ be a compact, star-shaped set containing the origin. For $r \geq 0$ the contact distribution function $H_B(r)$ of $\Xi$ is given by

$$1 - H_B(r) = \lambda \int_0^\infty t^{d-1} e^{-\lambda V_d(\cup_{y \in B \cap (0,tB)} b(y, r))} d\sigma_{d-1}(dt) dr.$$  

**Remark 1.** For $d = 2$ and $d = 3$, the spherical contact distribution function of $\Xi$ has been computed explicitly in Mache and Stoyan (1992). For $d = 2$, it is given by

$$H_s(r) = 1 - \frac{1}{2} e^{-4\pi \lambda r^2} + \frac{1}{2} e^{-6\pi \lambda r^2} - 2\pi \lambda \int_0^\infty \rho e^{-\lambda \left((\rho^2 + 2\rho r + r^2)^\frac{1}{2} + 6r \right)} d\rho,$$  

and for $d = 3$ we have

$$H_s(r) = 1 - 4\pi \lambda \int_0^r \rho^2 e^{-\lambda \frac{1}{2} \rho^2} (\rho^2 + \rho^2) d\rho$$  

**Remark 2.** The volume density $V_{V,d}$ of $\Xi_R$ equals the spherical contact distribution function of $\Xi$, i.e.,

$$V_{V,d}(\Xi_R) = H_s(R).$$  

For any random closed set $\Theta$ and a compact set $B \in \mathcal{B}$, the capacity functional is defined as $T_B(\Theta) = P(\Theta \cap B \neq \emptyset)$. It uniquely determines the distribution of $\Theta$ (Schneider and Weil, 2008, Theorem 2.1.3).

If $B \in \mathcal{X}$ is a convex body containing the origin, we have

$$T_{\Xi}(rB) = P(\Xi \cap rB \neq \emptyset) = 1 - P(rB \subset F_d(0)) = H_B(r)$$  

for any $r > 0$. The capacity functional of the dilated tessellation $\Xi_R$ is related to the contact distribution function of the original tessellation via

$$T_{\Xi}(rB) = 1 - P(rB \oplus b(0,R) \subset F_d(0)) = \frac{1}{2} H_B(b(0,R)) (r).$$  

**Theorem 2.** The contact distribution function $H_{B,R}$ of $\Xi_R$ is given by

$$1 - H_{B,R}(r) = \int_0^\infty \int_0^\infty t^{d-1} e^{-\lambda V_d(\cup_{y \in B \cap (0,tB)} b(y, r))} d\sigma_{d-1}(dt) dr.$$  

Proof. The application of the law of total probability to the definitions of $H_{B,R}$ and $T_{\Xi}(B)$ yields

$$H_{B,R}(r) = 1 - \frac{1 - T_{\Xi}(rB)}{1 - V_{V,d}(\Xi_R)}.$$  

Now use Eqs. 6, 5, and Theorem 1.

**Example 1** (Spherical contact distribution function). For $B = b(0,1)$, we have

$$T_{\Xi}(rB) = H_{b(0,1)}(r) = H_s(r + R).$$  

Hence, $H_{B,R}(r) = 1 - \frac{1 - H_s(r + R)}{1 - H_s(R)}$, where the explicit formulae for $H_s$ in $d = 2$ and $d = 3$ are given in Remark 1.

DENSITIES OF THE INTRINSIC VOLUMES

In this section we study the densities of the intrinsic volumes of the random closed set $\Xi_R$ in $\mathbb{R}^2$ and $\mathbb{R}^3$. These exist since $\Xi_R$ fulfills the above-mentioned conditions for the existence of the limit (Eq. 1) which follows from the representation (Eq. 2).

We derive integral formulae for the densities of area and boundary length in $\mathbb{R}^2$ and for the volume density and the specific surface area in $\mathbb{R}^3$. The interpretation of these two densities is straightforward and they are sufficient for the estimation of the parameters $\lambda$ and $R$.  

33
Theorem 3. For $d = 2$, the area density $A_{A,R}$ is given by $H_2(R)$ in Eq. 3. The density of the boundary length $L_{A,R}$ of $\mathcal{Z}_R$ is given by

$$L_{A,R} = 4\pi \lambda \left( R e^{-4\pi \lambda R^2} - R e^{-6\pi \lambda R^2} \right) \right) + 4\lambda \int_R^\infty \rho e^{-\lambda \left( (4\lambda^2 + 2\rho^2)(\pi - \arccos \frac{\rho}{R}) + 6R\sqrt{\rho^2 - R^2} \right)} \left( R \left( \pi - \arccos \frac{R}{\rho} \right) + \sqrt{\rho^2 - R^2} \right) d\rho \right).$$

Proof. By Stoyan et al. (1995, Equation (6.2.4)), $L_{A,R}$ can be obtained as $L_{A,R} = (1 - A_{A,R})H_2^R(0)$.\qed

The integrals can be evaluated numerically. Plots for the densities of area and boundary length for a tessellation with unit intensity as functions of $R$ are shown in Fig. 3.

Theorem 4. For $d = 3$, the volume density $V_{V,R}$ of $\mathcal{Z}_R$ is given by $H_3(R)$ in Eq. 4. The surface area density $S_{V,R}$ of $\mathcal{Z}_R$ is given by

$$S_{V,R} = \frac{64\pi^2 \lambda^2}{3} \left( \frac{R}{2} e^{-\lambda \left( \frac{32\pi}{3} R (R^2 + \rho^2) (3R^2 + \rho^2) d\rho \right) + \int_R^\infty \rho e^{-\lambda \left( \frac{4\pi}{3\rho} (R + \rho)^4 (R + \rho)^3 d\rho} \right).$$

Proof. By Stoyan et al. (1995, Equation (6.2.4)), $S_{V,R}$ can be obtained as $S_{V,R} = (1 - V_{V,R})H_3^R(0)$.\qed

As in the planar case, the integrals can be evaluated numerically. Plots for the densities of volume and surface area for a tessellation with unit intensity as functions of $R$ are shown in Fig. 4.

For the interpretation of the remaining intrinsic volumes we notice that the complement of the random closed set $\mathcal{Z}_R$ forms a particle process in $\mathbb{R}^d$. It consists of non-overlapping convex polytopes which are eroded versions of the cells of the tessellation (see Fig. 2). We denote its intensity by $N_{V,R}$. Alternatively, we can write $N_{V,R} = p_R \lambda$, where $p_R$ is the probability that the typical cell ‘survives’ an erosion with a ball of radius $R$. For particle processes $X$ of non-overlapping objects with intensity $N_V$, the additivity of the intrinsic volumes yields

$$V_{V,R}(X) = N_V EV_k(X_0),$$

where $X_0$ is the typical grain of $X$ (Schneider and Weil, 2008).

Let now $X \subset \mathbb{R}^d$ be a compact, polyconvex and topologically regular set. Then the Euler characteristics of $X$ and the topological closure $\overline{X}$ of the complement of $X$ are related via

$$\chi(X) = (-1)^{d+1} \chi(\overline{X}),$$

see (Ohser and Schladitz, 2009, Formula (3.11)). Since $\mathcal{Z}_R \cap K$ with $K \in \mathcal{K}$ fulfills the requirements listed above, the density of the Euler characteristic can be approximated as $(-1)^{d+1}N_{V,R}$.

For the interpretation of $M_{V,R}$, we remark that the densities of the integral of mean curvature of $X$ and $\overline{X}$ just differ by a factor $-1$ (Ohser and Schladitz, 2009, p.159). Furthermore, the mean width $\bar{b}(K)$ of a convex body $K \in \mathcal{K}$ is related to its integral of mean curvature $M(K)$ via $M(K) = 2\pi \bar{b}(K)$ (Ohser and Schladitz, 2009, p.21). Hence, we get

$$M_{V,R} = -2\pi N_{V,R} \bar{b}_R,$$

where $\bar{b}_R$ is the expected mean width of the eroded cells of the tessellation.

In applications, the values of $M_{V,R}$ and $\chi_{V,R}$ (or $\chi_{A,R}$) can be used to validate a Poisson-Voronoi model which has been fitted using $V_{V,R}$ and $S_{V,R}$ (or $A_{A,R}$ and $L_{A,R}$). This is of particular interest since the estimators for the intrinsic volume densities used in this paper estimate all densities simultaneously (see also Ohser and Schladitz (2009)). Hence, characteristics for parameter estimation and model validation can be obtained in one single estimation step.

For these reasons, we believe that also the intrinsic volume densities $\chi_{A,R}$ for $d = 2$ and $M_{V,R}$ and $\chi_{V,R}$ for $d = 3$ deserve an investigation. Since we do not see a way to obtain analytical formulae for them, they are studied by simulation.

For $d = 2$, the density of the Euler characteristic was estimated from 50 realisations of a stationary Poisson-Voronoi tessellation with intensity $\lambda = 1$ in a square with edge length 40. The dilated edge systems of the tessellations were discretized in binary images with $2000 \times 2000$ pixels. The density of the Euler characteristic was estimated from the image data using the estimator proposed in Ohser and Schladitz (2009) which is implemented in the MAVI software package (Fraunhofer ITWM, 2005); see Fig. 3 and Table 1.

In the case $d = 3$, the densities of the integral of mean curvature and the Euler characteristic were estimated from 30 realisations of Poisson-Voronoi tessellations with unit intensity in a cube with edge length 10, i.e., containing 1000 cells on average. The dilated facet systems of the tessellation were
discretized in binary images with $1000^3$ voxels. Again, the densities of the intrinsic volumes were estimated from the image data using the estimators proposed in Ohser and Schladitz (2009); see Fig. 4 and Table 2.

Fig. 3. Densities of area, boundary length, and Euler characteristic of a dilated Poisson-Voronoi tessellation in $\mathbb{R}^2$ with intensity $\lambda = 1$ as a function of $R$. Circles correspond to the mean values estimated from 50 realisations of the model. The envelopes plotted in grey are the minimal and maximal values from the 50 realisations.

Fig. 4. Densities of volume, surface area, integral of mean curvature, and Euler characteristic of a dilated Poisson-Voronoi tessellation in $\mathbb{R}^3$ with intensity $\lambda = 1$ as a function of $R$. Circles correspond to the mean values estimated from 30 realisations of the model. The envelopes plotted in grey are the minimal and maximal values from the 30 realisations.
Remark 3. The limits of the intrinsic volume densities for $\Xi_\mathbb{R}$ for $K \to 0$ can be computed from geometric characteristics of the Poisson-Voronoi tessellation. For $d = 2$, we have

$$A_{A,0} = 0,$$
$$L_{A,0} = 2L = 4\lambda^{\frac{1}{2}}, \text{ and}$$
$$\chi_{A,0} = -\lambda,$$

where $L$ denotes the expected total edge length of the Poisson Voronoi tessellation per unit volume. For $d = 3$, we get

$$V_{V,0} = 0,$$
$$S_{V,0} = 2S = 5.820\lambda^{\frac{1}{2}},$$
$$M_{V,0} = -2\pi \bar{b} \approx -9.161\lambda^{-\frac{1}{3}}, \text{ and}$$
$$\chi_{V,0} = \lambda,$$

where $S$ denotes the expected total surface area of the facets of the tessellation per unit volume, and $\bar{b}$ is the expected mean width of its typical cell.

APPLICATION

In this section we apply the formulae for $V_{V,R}$ and $S_{V,R}$ derived above to fit a Poisson-Voronoi model to the microstructure of a closed-cell polymer foam which is used for the insulation of buildings. We analysed a CT image of the material consisting of 480 $\times$ 480 $\times$ 360 voxels with a voxel edge length of 10.21 $\mu$m. The densities of volume and surface area were estimated from a binarised version of the image using the estimators of Ohser and Schladitz (2009). The estimated values are $V_V = 7.91\%$ and $S_V = 6.983$ mm$^{-1}$. Based on the formulae given in Theorem 4 we fitted the parameters of the dilated Poisson-Voronoi tessellation to the estimated values using a Nelder-Mead simplex optimisation procedure (Nelder and Mead, 1965). The distance function which was minimised was

$$f(\lambda, R) = \left(\frac{\hat{V}_V - V_V^p(\lambda, R)}{\hat{V}_V}\right)^2 + \left(\frac{\hat{S}_V - S_V^p(\lambda, R)}{\hat{S}_V}\right)^2,$$

where $V_V^p(\lambda, R)$ and $S_V^p(\lambda, R)$ are the characteristics for a dilated Poisson-Voronoi tessellation with intensity $\lambda$ and dilation radius $R$. The results are $\hat{\lambda} = 1.989$ mm$^{-3}$ and $\hat{R} = 9.904$ $\mu$m which corresponds to $V_V^p(\hat{\lambda}, \hat{R}) = 7.08\%$ and $S_V^p(\hat{\lambda}, \hat{R}) = 6.984$ mm$^{-1}$. The fit of these values is relatively close. However, the visualisations of the original sample and the model shown in Fig. 5 show a certain deviation in the cell shape. This indicates that a tessellation with more regular cell shapes might be a preferable model for this material.

Unfortunately, the validation method suggested above is not applicable for this data set: it is well-known that the estimators for $M_V$ and $\chi_V$ are sensitive to the resolution of the image (Ohser and Schladitz, 2009). In the current image, the resolution is not high enough to fully resolve the cell walls which are slightly perforated in the binary image. While the effect on the estimates of the volume and surface density (loss of several pixels) is negligible, the estimates $\hat{M}_V = 10.61$ mm$^{-2}$ and $\hat{\chi}_V = -401.2$ mm$^{-3}$ are highly affected. In particular, the discussion in the previous section shows that a negative value of $M_V$ should be expected for the dilated facet system of a three-dimensional tessellation while $\chi_V$ should be positive. In the foam example, the signs of both characteristics are reversed. Consequently, their interpretation is questionable and they cannot be used for model validation.

Fig. 5. Visualisations and sectional images of the original foam (left) and the fitted Poisson-Voronoi model (right). Sample and CT imaging: R. Schlümp, Fraunhofer IMW Halle.

Remark 4. If $\Xi$ is a stationary and isotropic random closed set in $\mathbb{R}^3$ then its volume and surface area density are related to the area density $A_\lambda$ and the boundary length $L_\lambda$ of the two-dimensional section $\Xi \cap (\mathbb{R}^2 \times \{0\})$ via

$$V_V = A_\lambda^S \quad \text{and} \quad S_V = \frac{4}{\pi} L_\lambda^S.$$
Hence, a three-dimensional model can also be fitted based on measurements in 2D images. Whenever possible, three-dimensional data should be preferred for the model fit. In the data presented above, the area fractions estimated from sectional images vary between 6.74% and 9.41%. Hence, the choice of a single slice might lead to a significantly different estimation of the model parameters.

Nevertheless, the possibility to fit a model using 2D observations is valuable for materials which are not tractable to 3D imaging techniques, e.g., the sinter material shown in Fig. 1.

**DISCUSSION**

We have studied the densities of the intrinsic volumes of the dilated facet systems of stationary Poisson-Voronoi tessellations in $\mathbb{R}^2$ and $\mathbb{R}^3$. We obtained analytical formulae for the densities of the area and boundary length in $\mathbb{R}^2$ and the volume and surface area in $\mathbb{R}^3$. The densities of the remaining intrinsic volumes could only be studied by simulation. The results presented in this paper allow for a quick and easy fitting of stationary Poisson-Voronoi tessellations to cellular materials. Since two of the intrinsic volume densities are sufficient for the parameter estimation, the remaining ones can be used for model validation.

In many applications, tessellation models generated by regular point processes might provide better model fitting results. Nevertheless, the Poisson-Voronoi tessellation could be sufficient in cases where the cell shape is of minor importance. Furthermore, thanks to its analytical tractability it plays an important role as reference model. Due to the independent placement of the cells’ generators in the Poisson-Voronoi tessellation this model yields a very disordered structure. Hence, it could also be seen as an extreme case which can possibly provide bounds for characteristics of more regular tessellation models.

Formulae for the spherical contact distribution function of more general Voronoi tessellations and of Poisson Laguerre tessellations were given in Heinrich (1998) and Lautensack (2007), respectively. These formulae could be used to extend the results from this paper to further tessellation models. So far, however, these formulae are given in less explicit form than the formulae for the Poisson-Voronoi tessellation which hinders their applicability.

**APPENDIX**

Tables 1 and 2 contain the means and standard deviations of $\chi_{A,R}$, $M_{V,R}$, and $\chi_{V,R}$ estimated by simulation.

Table 1. Values of $\chi_{A,R}$ in $\mathbb{R}^2$ estimated by simulation.

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</tr>
<tr>
<td>0.650</td>
<td>-0.0406</td>
<td>0.0044</td>
</tr>
</tbody>
</table>

Table 2. Values of $M_{V,R}$ and $\chi_{V,R}$ in $\mathbb{R}^3$ estimated by simulation.

<table>
<thead>
<tr>
<th>$R$</th>
<th>$M_{V,R}$ mean</th>
<th>$M_{V,R}$ sd</th>
<th>$\chi_{V,R}$ mean</th>
<th>$\chi_{V,R}$ sd</th>
</tr>
</thead>
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<tr>
<td>0.025</td>
<td>-8.9695</td>
<td>0.2039</td>
<td>1.0025</td>
<td>0.0335</td>
</tr>
<tr>
<td>0.050</td>
<td>-8.5529</td>
<td>0.1901</td>
<td>1.0022</td>
<td>0.0347</td>
</tr>
<tr>
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<td>-7.7027</td>
<td>0.1598</td>
<td>1.0026</td>
<td>0.0341</td>
</tr>
<tr>
<td>0.150</td>
<td>-6.8224</td>
<td>0.1268</td>
<td>1.0014</td>
<td>0.0348</td>
</tr>
<tr>
<td>0.200</td>
<td>-5.8910</td>
<td>0.0928</td>
<td>0.9989</td>
<td>0.0329</td>
</tr>
<tr>
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<td>0.0550</td>
<td>0.9758</td>
<td>0.0327</td>
</tr>
<tr>
<td>0.300</td>
<td>-3.7635</td>
<td>0.0250</td>
<td>0.9218</td>
<td>0.0283</td>
</tr>
<tr>
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<td>0.0390</td>
<td>0.7827</td>
<td>0.0193</td>
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<tr>
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<td>0.0516</td>
<td>0.5707</td>
<td>0.0141</td>
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<td>0.0161</td>
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<tr>
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<td>0.0138</td>
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<td>0.0532</td>
<td>0.0069</td>
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<tr>
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<td>-0.0199</td>
<td>0.0082</td>
<td>0.0134</td>
<td>0.0043</td>
</tr>
</tbody>
</table>

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REFERENCES


