

Voronoi-based estimation of Minkowski tensors from digital images

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Joint work with Daniel Hug and Markus Kiderlen

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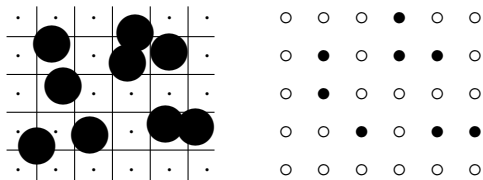
CSGB Center for Stochastic Geometry and Advanced Bioimaging

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Digital images

Suppose we study an object $X \subseteq \mathbb{R}^d$. (For instance via a microscope or scanner.)

The only information we have about X is a (black-and-white) digital image:

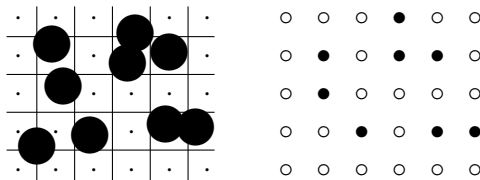


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The pixel midpoints form a lattice \mathbb{L} . ($\mathbb{L} = \mathbb{Z}^d$)

Mathematically, the information in a black-and-white image is the set

$$X \cap \mathbb{L}$$

of black pixel midpoints.

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we want to derive information about the geometry of X .

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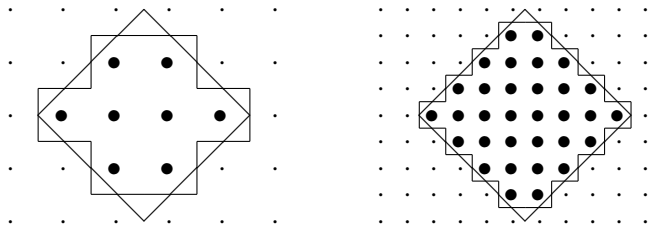
Change of resolution corresponds to scaling \mathbb{L} by some $a > 0$. We then have the information

$$X \cap a\mathbb{L}.$$

Want estimators to converge when $a \rightarrow 0$.

Naive approach

Approximate X by the union of black pixels.



Good approximation of volume.

Boundary approximation is generally poor.

The boundary length is approximately $\sqrt{2}$ times too large.

Conditions on the object

We assume that our object $X \subseteq \mathbb{R}^d$

- is compact.
- is topologically regular, i.e. $X = \overline{\text{int}(X)}$.
- has positive reach, i.e. $\text{Reach}(X) > 0$.

Definition

Let $\text{Reach}(X)$ be the largest number such that all $x \in \mathbb{R}^d$ with $d(X, x) < \text{Reach}(X)$ has a unique closest point in X .

Convex sets and C^2 manifolds have positive reach.

Minkowski volume tensors

For $r, s \geq 0$, define the r -tensor:

$$\Phi_d^{r,0}(X) = \frac{1}{r!} \int_X x^r dx$$

In particular, $\Phi_d^{0,0}(X)$ is volume.

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Can estimate this by a Riemann sum

$$\Phi_d^{r,0}(X) \approx a^d c_{\mathbb{L}} \frac{1}{r!} \sum_{x \in X \cap a\mathbb{L}} x^r$$

where $c_{\mathbb{L}}$ is the volume of a lattice cell in \mathbb{L} .

General Minkowski tensors

For $r, s \geq 0$ and $k = 0, \dots, d - 1$, define

$$\Phi_k^{r,s}(X) = c_{r,s,k} \int_{\Sigma} x^r u^s C_k(X; d(x, u))$$

where:

$$\Sigma = \mathbb{R}^d \times \mathcal{S}^{d-1},$$

$C_k(X; \cdot)$ is the k 'th generalized curvature measure on Σ ,
 $x^r u^s$ means the symmetric tensor product.

Again, $\Phi_k^{0,0}(X)$ is the k th the intrinsic volume.

The Steiner formula

Let $0 \leq R$ and define the parallel set

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$$\mathcal{V}_R^{r,s}(X) = \int_{X^R} p_X(x)^r (x - p_X(x))^s dx.$$

$\mathcal{V}_R^{0,2}(X)$ is the (total) Voronoi covariance measure of *Mérogot et al.* (2010).

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When $R < \text{Reach}(X)$, the (generalized) Steiner formula yields

$$\mathcal{V}_R^{r,s}(X) = c_{r,s} \sum_{k=0}^d \kappa_{k+s} R^{s+k} \Phi_{d-k}^{r,s}(X). \quad (1)$$

κ_k is the volume of the unit ball in \mathbb{R}^k .

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For $r = s = 0$, this is the usual Steiner formula.

Idea of estimation

Choose $0 < R_0 < \dots < R_d < \text{Reach}(X)$.

Write the equations (1) in matrix form:

$$\begin{pmatrix} \mathcal{V}_{R_0}^{r,s}(X) \\ \vdots \\ \mathcal{V}_{R_d}^{r,s}(X) \end{pmatrix} = C_{r,s} \begin{pmatrix} \kappa_s R_0^s & \dots & \kappa_{s+d} R_0^{s+d} \\ \vdots & & \vdots \\ \kappa_s R_d^s & \dots & \kappa_{s+d} R_d^{s+d} \end{pmatrix} \begin{pmatrix} \Phi_d^{r,s}(X) \\ \vdots \\ \Phi_0^{r,s}(X) \end{pmatrix}$$

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Can solve for the Minkowski tensors:

$$\begin{pmatrix} \Phi_d^{r,s}(X) \\ \vdots \\ \Phi_0^{r,s}(X) \end{pmatrix} = \frac{1}{C_{r,s}} \begin{pmatrix} \kappa_s R_0^s & \dots & \kappa_{s+d} R_0^{s+d} \\ \vdots & & \vdots \\ \kappa_s R_d^s & \dots & \kappa_{s+d} R_d^{s+d} \end{pmatrix}^{-1} \begin{pmatrix} \mathcal{V}_{R_0}^{r,s}(X) \\ \vdots \\ \mathcal{V}_{R_d}^{r,s}(X) \end{pmatrix}$$

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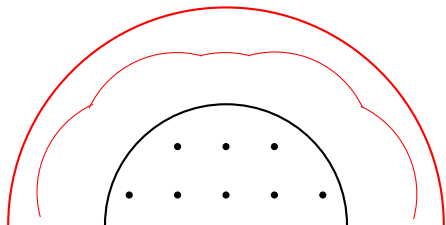
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Idea: approximate $\mathcal{V}_{R_i}^{r,s}(X)$ by $\mathcal{V}_{R_i}^{r,s}(X \cap a\mathbb{L})$.

The algorithm

By definition

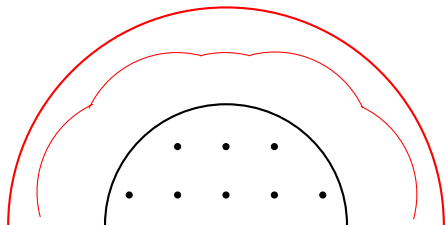
$$\mathcal{V}_R^{r,s}(X \cap a\mathbb{L}) = \int_{(X \cap a\mathbb{L})^R} p_{X \cap a\mathbb{L}}(y)^r (y - p_{X \cap a\mathbb{L}}(y))^s dy.$$



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By definition

$$\mathcal{V}_R^{r,s}(X \cap a\mathbb{L}) = \int_{(X \cap a\mathbb{L})^R} p_{X \cap a\mathbb{L}}(y)^r (y - p_{X \cap a\mathbb{L}}(y))^s dy.$$



For intrinsic volumes:

$$\begin{aligned}\mathcal{V}_R^{0,0}(X) &= V_d(X^R), \\ \mathcal{V}_R^{0,0}(X \cap a\mathbb{L}) &= V_d((X \cap a\mathbb{L})^R).\end{aligned}$$

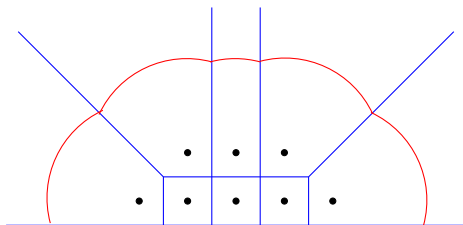
Voronoi expression

For each $x \in X \cap a\mathbb{L}$, define the Voronoi cell of x by

$$V_x = \{y \in \mathbb{R}^d \mid \forall z \in X \cap a\mathbb{L} : |y - x| \leq |y - z|\}.$$

Then $(X \cap a\mathbb{L})^R = \bigcup_{x \in X \cap a\mathbb{L}} V_x \cap B(x, R)$ and hence

$$\begin{aligned} \mathcal{V}_R^{r,s}(X \cap a\mathbb{L}) &= \sum_{x \in X \cap a\mathbb{L}} \int_{V_x \cap B(x,R)} x^r (y-x)^s dy \\ &= \sum_{x \in X \cap a\mathbb{L}} x^r \int_{(V_x - x) \cap B(0,R)} z^s dz. \end{aligned}$$



Convergence

Would like our estimators to converge to the true value when $a \rightarrow 0$.

Recall that:

$$\begin{pmatrix} \Phi_d^{r,s}(X) \\ \vdots \\ \Phi_0^{r,s}(X) \end{pmatrix} = \frac{1}{c_{r,s}} \begin{pmatrix} \kappa_s R_0^s & \dots & \kappa_{s+d} R_0^{s+d} \\ \vdots & & \vdots \\ \kappa_s R_d^s & \dots & \kappa_{s+d} R_d^{s+d} \end{pmatrix}^{-1} \begin{pmatrix} \mathcal{V}_{R_0}^{r,s}(X) \\ \vdots \\ \mathcal{V}_{R_d}^{r,s}(X) \end{pmatrix}$$

So it is enough that

$$\lim_{a \rightarrow 0} \mathcal{V}_R^{r,s}(X \cap a\mathbb{L}) = \mathcal{V}_R^{r,s}(X).$$

Convergence - intrinsic volumes

For $X_1, X_2 \subseteq \mathbb{R}^d$ compact, define the Hausdorff distance

$$d_H(X_1, X_2) = \inf\{\varepsilon > 0 \mid X_1 \subseteq X_2^\varepsilon, X_2 \subseteq X_1^\varepsilon\}.$$

Theorem (Chazal, Cohen-Steiner, Mérigot (2010))

Let $X_1, X_2 \subseteq \mathbb{R}^d$ compact. Let $d_H(X_1, X_2) < \frac{R}{2}$. Then

$$|V_d(X_1^R) - V_d(X_2^R)| \leq C(d, X_1, R)d_H(X_1, X_2).$$

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Corollary

$$\lim_{a \rightarrow 0} \mathcal{V}_R^{0,0}(X \cap a\mathbb{L}) = \mathcal{V}_R^{0,0}(X).$$

Convergence - general tensors

Theorem (Merigot et al. (2010), Hug, Kiderlen, S. (2014))

Let $X_1, X_2 \subseteq \mathbb{R}^d$ be compact. Assume $d_H(X_1, X_2) < \min\{\frac{R}{2}, \text{diam}(X_1), \frac{\text{diam}(X_1)^2}{R - d_H(X_1, X_2)}\}$. Then

$$|\mathcal{V}_R^{r,s}(X_1) - \mathcal{V}_R^{r,s}(X_2)| \leq C(d, r, s, X_1, R) d_H(X_1, X_2)^{\frac{1}{2}}.$$

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Corollary

$$\lim_{a \rightarrow 0} \mathcal{V}_R^{r,s}(X \cap a\mathbb{L}) = \mathcal{V}_R^{r,s}(X).$$

If X is convex or a C^2 manifold, then the convergence speed is $O(\sqrt{a})$.

A few words about the proof

It is enough to show that for a basis $e_1, \dots, e_d \in \mathbb{R}^d$, the evaluations satisfy

$$\begin{aligned} & |\mathcal{V}_R^{r,s}(X_1)(e_{i_1}, \dots, e_{i_{r+s}}) - \mathcal{V}_R^{r,s}(X_2)(e_{i_1}, \dots, e_{i_{r+s}})| \\ & \leq \tilde{C}(d, r, s, X_1, R) d_H(X_1, X_2)^{\frac{1}{2}}. \end{aligned}$$

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The left hand side can be written as

$$\left| \int_{X_1^R} f(p_{X_1}(x), x - p_{X_1}(x)) dx - \int_{X_2^R} f(p_{X_2}(x), x - p_{X_2}(x)) dx \right|.$$

where $f : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ is some locally Lipschitz function.

The key ingredient is now a theorem by Chazal, Cohen-Steiner, and Mérigot:

$$\int_E |p_{X_1}(x) - p_{X_2}(x)| dx \leq C(d, X_1, E) d_H(X_1, X_2)^{\frac{1}{2}}.$$

Local tensors

We can consider the tensor valued measure, given on a Borel set $A \subseteq \mathbb{R}^d$ by

$$\Phi_k^{r,s}(X; A) = c_{r,s,k} \int_{A \times S^{d-1}} x^r u^s C_k(X; d(x, u)).$$

Define the Voronoi tensor measure

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Theorem

Let $X_i, X \subseteq \mathbb{R}^d$ be compact sets. If $\lim_{i \rightarrow \infty} d_H(X_i, X) = 0$, then

$$\lim_{i \rightarrow \infty} \mathcal{V}_R^{r,s}(X_i; A) = \mathcal{V}_R^{r,s}(X; A)$$

for every Borel set A that satisfies $\mathcal{H}^d(p_X^{-1}(\partial A) \cap X^R) = 0$.

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Local tensors with $A \subseteq \Sigma$?

Conclusions:

- Get algorithm for estimation of all Minkowski tensors.
- Proof of convergence when resolution goes to infinity.
- Simple expression in terms of Voronoi cells.
- Slower than local algorithms.
- Applies to other approximations of X than digital images.

Discussion

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Future directions:

- Performance in practice?
- How to choose R_i ?
- Extension to polyconvex sets?
- Extension to grey-valued images?