

Spatial modeling of micro-structure in porous material

Smögen Workshop, August 25-29, 2014

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Joint with AstraZeneca R&D Mölndal and Fraunhofer ITWM

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Motivation and data

- Conclusions prior to modeling

Germ-Grain models

- Boolean model

- Quermass-interaction model

Multi-layer modeling of pore connectivity

- Example from the literature

Discussion

Summary

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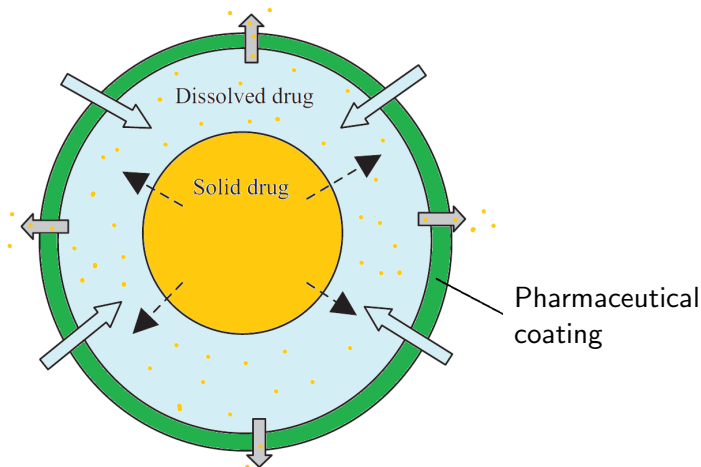
Goal

Characterizing and **modeling** micro-structures of porous materials
in order to better understand and control their properties

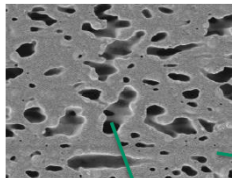


Taking germ-grain models
to the next level!?

Motivation: Controlled drug release

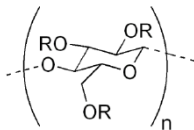


Polymer blended coating



Hydroxypropyl
cellulose (HPC)

$R=H$ or $CH_2CH(OH)CH_3$



Ethyl cellulose
(EC)

$R=H$ or CH_2CH_3

Summary: Important data features

- From images
 - isotropic surface, but **anisotropic** cross-section
- Results from experiments
 - film is permeable → percolation path through sample
 - pore connectivity is essential
- Results from 2D image analysis
 - HPC area fraction and **pore shape** significant factors for permeability
 - number of endpoints in skeleton good measure of pore shape

Conclusions prior to modeling

3D modeling with marked point processes

Step I Germ-grain models for the surfaces

- stack of marked point processes

$$\Phi = \{[X_i, M_i] : i = 1, 2, \dots\}$$

- assign shape parameter(s) to points as marks M_i defining the grains

Step II Link surfaces by displacement vectors capturing the chain like HPC structure in z-direction

- skeleton
- mark correlation

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Germ-Grain models

Based on marked point process $\Phi_M = \{[X_i, M_i] : i = 1, 2, \dots\}$

$$\Xi = \bigcup_{i=1}^{\infty} (\underbrace{G_{M_i}}_{\text{grain}} + \underbrace{X_i}_{\text{germ}})$$

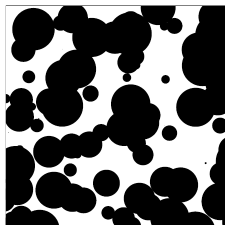


- $\Phi = \{X_i\}$ point process
- G_{M_i} (random) compact set defined by parameter(s) M_i
- $(G_{M_i} + X_i) = \{x + X_i : x \in G_{M_i}\}$

Boolean model

Overlapping sphere system $\Xi = \bigcup_{i=1}^{\infty} (\underbrace{G_{M_i}}_{\text{grain}} + \underbrace{X_i}_{\text{germ}})$

- $\{X_i\}$ (in)homogeneous Poisson process
- $G_{M_i} = b(o, M_i)$ with fixed or random radii
 $M_i : \Omega \rightarrow (0, \infty)$
- + Well known properties
- Limited variety of generated structures



Quermass-interaction model

Adding morphological interaction between grains s.t.
absolutely continuous w.r.t. marked Poisson process with density
 $\propto \exp(-H_{\Phi}^{\theta})$

$$H_{\Phi_W}^{\theta} = \theta_1 \mathcal{A}(\Phi_W) + \theta_2 \mathcal{L}(\Phi_W) + \theta_3 \chi(\Phi_W)$$

$\theta = (\theta_1, \theta_2, \theta_3) \in \mathbb{R}^3$ for Quermass functionals

\mathcal{A} area

\mathcal{L} perimeter

χ Euler number (# objects - # wholes)

$\Phi_W = \{X_i\}$ Poisson process in observation window $W \in \mathbb{R}^d$

Quermass-interaction model: Examples

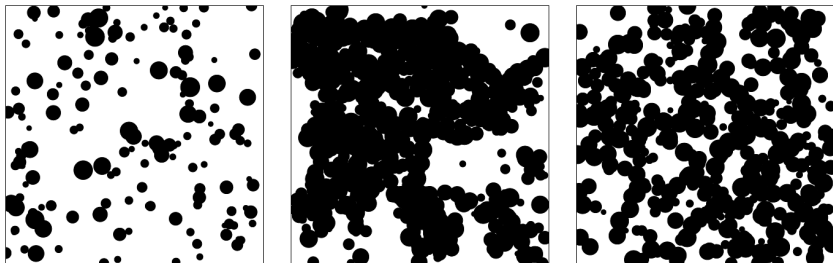


Figure : Examples based on marked Poisson process with intensity λ and spheres with radii $M_i \sim U[0.5, 2]$: left $\lambda = 0.1$, $\theta = (0.2, 0, 0)$, middle $\lambda = 0.2$, $\theta = (0, 0.4, 0)$, right $\lambda = 0.1$, $\theta = (0, 0, 1)$

[DLS14]

Quermass-interaction model

Adding morphological (Quermass) interaction between grains via

$$H_{\Phi_W}^\theta = \theta_1 \mathcal{A}(\Phi_W) + \theta_2 \mathcal{L}(\Phi_W) + \theta_3 \chi(\Phi_W)$$

- + Large variety of generated structures
- Complicated parameter estimation based on K chosen test functions f_k for Takacs-Fiksel method

$$(\hat{\lambda}_\Phi, \hat{\theta}) := \arg \min_{\lambda_\Phi, \theta} \sum_{k=1}^K \underbrace{(C_W^{\lambda_\Phi, \theta}(\Phi; f_k))^2}_{r.v.}$$

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Multi-layer modeling of pore connectivity

Task: Capture heterogeneity and correlation of chain like HPC

- Continuum percolation
 - Boolean model with dependent growth of radii: spheres grow until they have touched k neighbors
 - random-connection model based on a connection function

Multi-layer modeling of pore connectivity

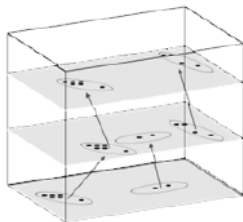
Task: Capture heterogeneity and correlation of chain like HPC

- Continuum percolation
 - Boolean model with dependent growth of radii: spheres grow until they have touched k neighbors
 - random-connection model based on a connection function
- **Markov chain** as stationary process in 3D based on displacement vectors as well as birth and death rates

Example: Matérn process Markov chain

Spatial birth-and-death process with random displacement¹

- Stack of $s \geq 1$ elliptical Matérn cluster processes $\{S_n^{(s)}, s \geq 1\}$ in \mathbb{R}^2
- High correlation of center locations in z-direction
 - stationary Markov chain with small displacements between
 - $0 < r_{min} < r_{max} \in \mathbb{R}$ of clusters
 - births and deaths of clusters



¹[SHT⁺11]

Example: Matérn process Markov chain ctd.

Underlying Poisson process $\{\Phi_n^{(s)}, n \geq 1\}$ with intensity λ defined by

- **Births:** $\{B_n^{(s)}\}$ Poisson process with intensity λ_B
- **Deaths:** $\{\delta_n^{(s)}\}$ *i.i.Ber*(p) s.t. $\lambda p + \lambda_B = \lambda$, independent of $\{B_n^{(s)}\}$
- **Displacement vectors:** $\{D_n^{(s)}\}$ *i.i.U* on $b(o, r_{max}) \setminus b(o, r_{min})$, independent of $\{B_n^{(s)}\}$ and $\{\delta_n^{(s)}\}$

$$\{\Phi_n^{(s+1)}\} = \bigcup_{i:\delta_i^{(s)}=1} \{\Phi_j^{(s)} + D_j^{(s)}\} \cup \{B_n^{(s)}\}$$

Example: Matérn process Markov chain ctd.

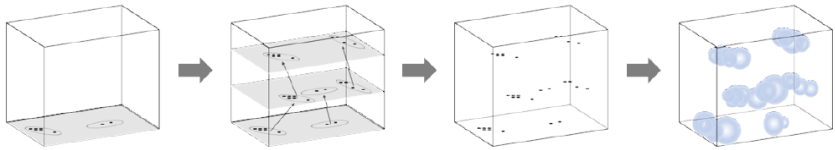


Figure : Modeling germs with Markov chain of Matérn cluster processes and spherical grains with random radii

[SHT⁺11]

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Main challenge: Germ-grain model

Question: What kind of grains should we model?

- Overlapping sphere system
 - Poisson process
 - Matérn cluster process
 - Gibbs process
 - Quermass-interaction model
- } — limited structure
 — fitting unobserved germs
- tricky parameter estimation

Main challenge: Germ-grain model

Question: What kind of grains should we model?

- Overlapping sphere system
 - Poisson process
 - Matérn cluster process
 - Gibbs process
 } — limited structure
 — fitting unobserved germs
- Quermass-interaction model — tricky parameter estimation
- System of overlapping compact but **more complex sets**
 - + better control over pore shape
 - + germs as midpoints of Watershed objects
 - unclear how to capture pore connectivity

More complex sets as grains

Possible candidates

- Watershed
- skeleton endpoints
- Fourier transform
- tensors
- set of basic shapes
- ...



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- **Task:** 3D model of porous micro-structure
 - pore shape and connectivity VERY important
- **Step I:** Extraction of marked point pattern
 - overlapping sphere system
 - system of overlapping more complex sets
- **Step II:** Connectivity in z-direction
 - continuum percolation
 - Markov chain
- **Question:** Use none of the above and base model on, e.g., a dilated skeleton?

Important references

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Thank you for your attention
and let us have a nice
discussion!