# Generation of 2-D Porous Media for a Study of Macroscopic Properties

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## Overview

- Why are porous media interesting?
- What is morphological information and what type of morphological information do we study?
- How do we reconstruct 2-D porous media?
- What insight will studies of the 2-D media give us with respect to macroscopic properties?

# **Porous Media**

- Any solid possessing pore space
  - Foam, soil, bones, and concrete
- We look at a 2-D lattice that consists of two phases
- What can we learn from studying the structure of these materials?



# **Morphological Information**

- Pertains to the physical shape, form, or structure of a substance
- Different measurements include the twopoint auto-correlation function, the linealpath function, the chord-length distribution function and the Minkowski functionals.

# **Minkowski Functionals**

 For a two-phase material in 2-D, describe the area and Euler characteristic of a particular phase, as well as the boundary length between the two phases





## **Iterative Calculation of Functionals**

- Define some base structure to calculate Minkowski functionals using an additive method
  Correct for over-
- counting and normalize



## Example – Minkowski Functionals

 Totals are consistent with values obtained by visual inspection



## Reconstructing Two-Phase Porous Media

- Cannot directly generate samples
- Use MCMC method to indirectly generate samples with desired values
  - Gibbs sampler
  - Metropolis-Hastings algorithm

# **Gibbs Sampler**

- Produces sample indirectly when generations from conditional distributions are possible
- Advantages
  - easy to implement
  - convergence guaranteed
- Disadvantage
  - conditional distributions are not always available

# **The Specifics**

- Want to sample from all structures consistent with a set of functionals
- Consider  $\pi(\vec{S})$  and

$$\pi_i(S_i) = \pi(S_i \mid \vec{S}_{-i})$$

#### Assume independence of 2-by-2 structures

 Probability of a point switching to phase one given S<sub>j</sub>, S<sub>k</sub> is

$$\Pr(p_{r,c} = 1 | S_j, S_k) = \frac{S_j}{S_j + S_k}$$



# **Results of Gibbs Sampler**

- Able to reproduce functionals for a random lattice because of independence of lattice points
- Cannot reproduce functionals for a lattice that is not random because of correlations between lattice points



# **Metropolis-Hastings Algorithm**

- Useful when non-iterative method of generating a sample from a distribution is not possible
- Modeled after physically cooling a substance until it is frozen into a state
- Advantages
  - Proven to work well
  - Fairly easy to implement
- Disadvantages
  - Convergence is not guaranteed
  - Many initial parameters

# **The Specifics**

• 
$$E = \sum_{i=0}^{2} (v_i^t - v_i^0)^2$$

• 
$$\Delta E = E' - E$$

•  $T = T_0 \left(\frac{T_f}{T_0}\right)^{(i/N)}$ 

• 
$$\Pr(\Delta E) = \begin{cases} 1, & \Delta E \le 0 \\ \exp(-\Delta E/T), & \Delta E > 0 \end{cases}$$

## Results of Metropolis-Hastings Algorithm

 Works for all possible Minkowski functionals regardless of correlations





## **Macroscopic Properties**

 Can macroscopic properties be determined from limited morphological information?



## Conclusion

- Metropolis-Hastings algorithm is effective for producing two-phase porous media in 2-D
- The generation of many samples with an array of values for the Minkowski functionals will enable the evaluation of the hypothesis driving this work

#### **References and Acknowledgments**

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