

Kinetic (Lattice) Monte Carlo for Helical Molecules and Magnetic Substrates

Jeffrey Kelling

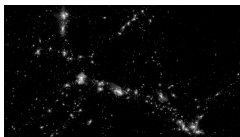
13th November 2020



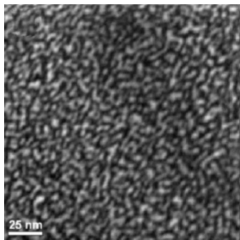
hZDR

Applications for Monte Carlo: Stochastic Processes

Phase Separation



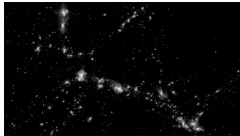
IllustrisTNG



Müller, T., Heinig, K.-H. et al. *Appl. Phys. Lett.* **85** 2373 (2004)

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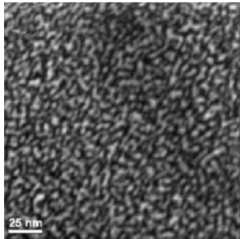


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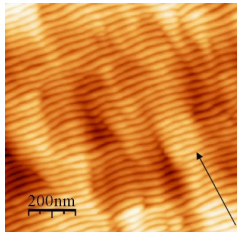
Driven Systems



http://en.wikipedia.org/wiki/File:Rub_al_Khali_002.JPG



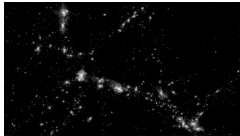
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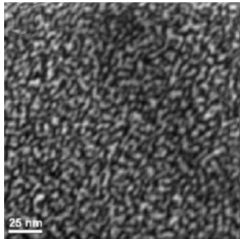
<https://www.hzdr.de/db/Cms?pOid=24344&pNid=2707>

Applications for Monte Carlo: Stochastic Processes

Phase Separation



IllustrisTNG

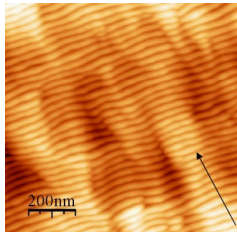


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Driven Systems



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<https://www.hzdr.de/db/Cms?pOid=24344&pNid=2707>

■ game theory
e. g.: Perc, Matjaž *Eur. J. Phys.*
38(4) 045801 (2017)

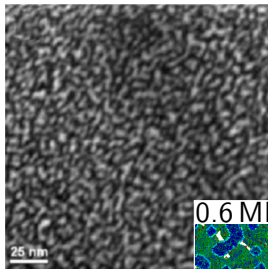
- sociology
- finance
- ...

- 1 Kinetic Metropolis Lattice Monte Carlo (KMC)
 - Multi-Surface Simulations
- 2 Example I: Phase separation in inhomogeneous systems
- 3 Example II: Structure formation under ion irradiation
- 4 Modeling in B2

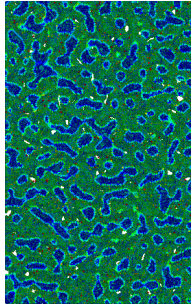
Kinetic Metropolis Lattice Monte Carlo (KMC)

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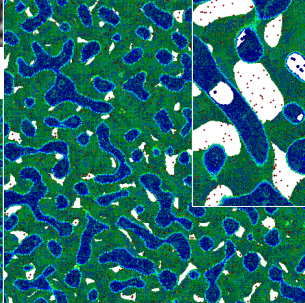
Simulation of Nano Structure Self-Organization



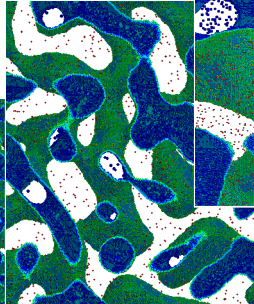
0.1 MMCS



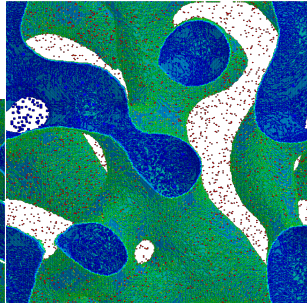
0.6 MMCS



20.5 MMCS

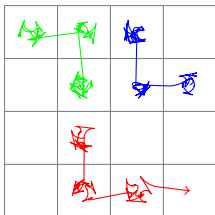


150.05 MMCS



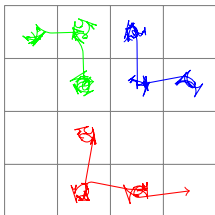
bulk particle
interface particle
dissolved particle
empty space:
embedding matrix

Molecular Dynamics



movement of impurity atoms

Molecular Dynamics



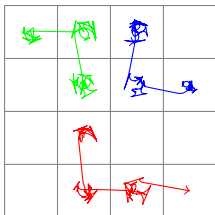
movement of impurity atoms

solve Hamilton's equations

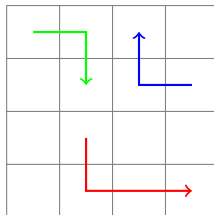
$$\dot{p}_i = -\frac{\partial H}{\partial q_i} \text{ and } \dot{q}_i = \frac{\partial H}{\partial p_i}$$

⇒ “nature’s random number generator”

Molecular Dynamics → Cellular Automaton



movement of impurity atoms



on-lattice movement of impurity atoms

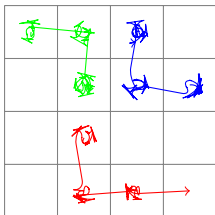
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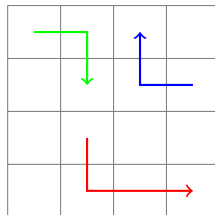
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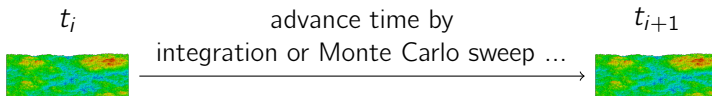
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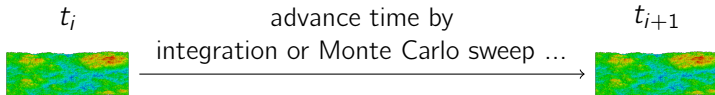
- we know how, from *thermodynamics*

⇒ stochastic process defined by Metropolis algorithm^a

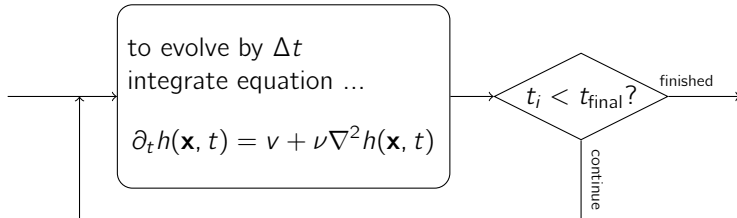
^aStrobel, Heing, Möller, *Phys. Rev. B* **64** 245422 (2001)



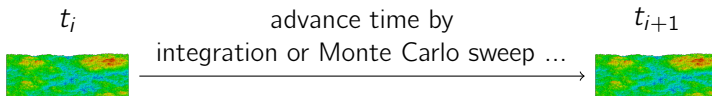
- simulated physical system evolves in discrete time steps Δt



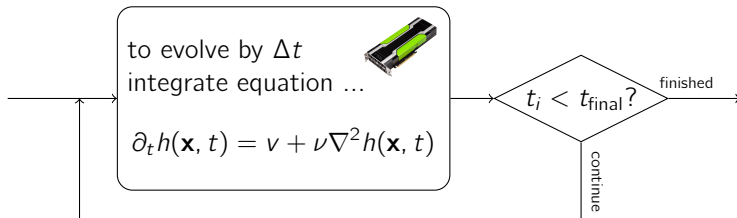
Phase Field Methods



- simulated physical system evolves in discrete time steps Δt
- more accurate the smaller Δt and the larger the system

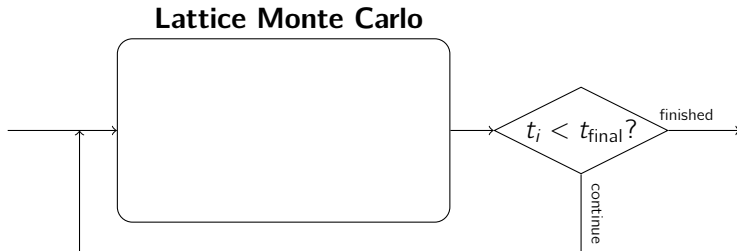
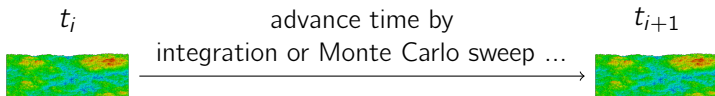


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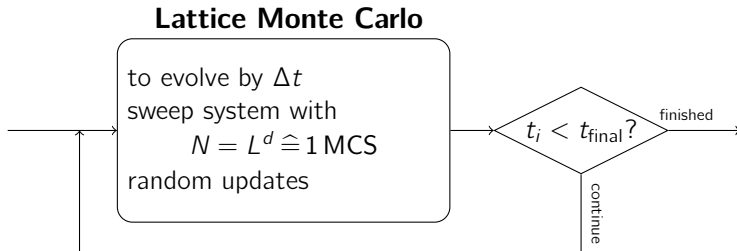
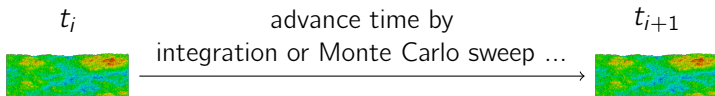


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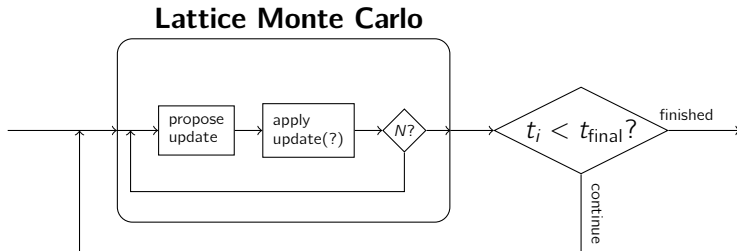
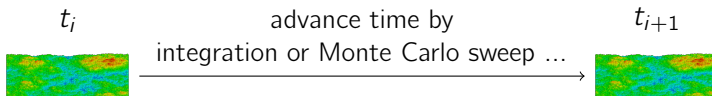
Simulation



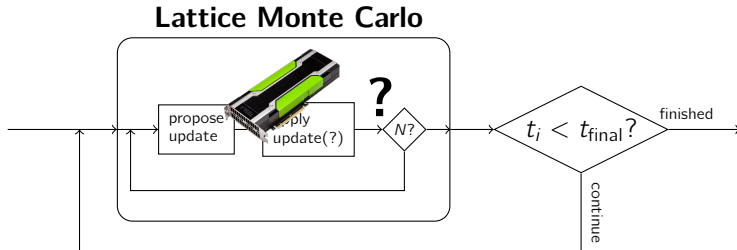
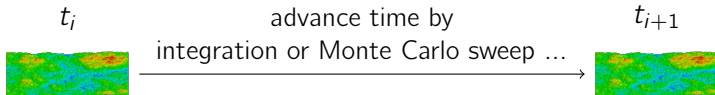
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- simulated physical system evolves in discrete time steps Δt
- more accurate the larger the system
- updates are simple, but need to be applied in order
- uncorrelated updates

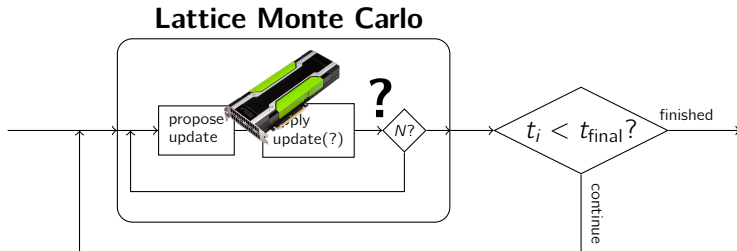
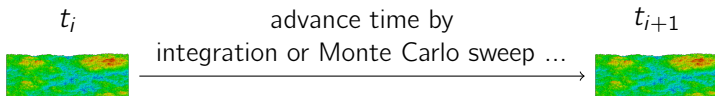


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Simulation



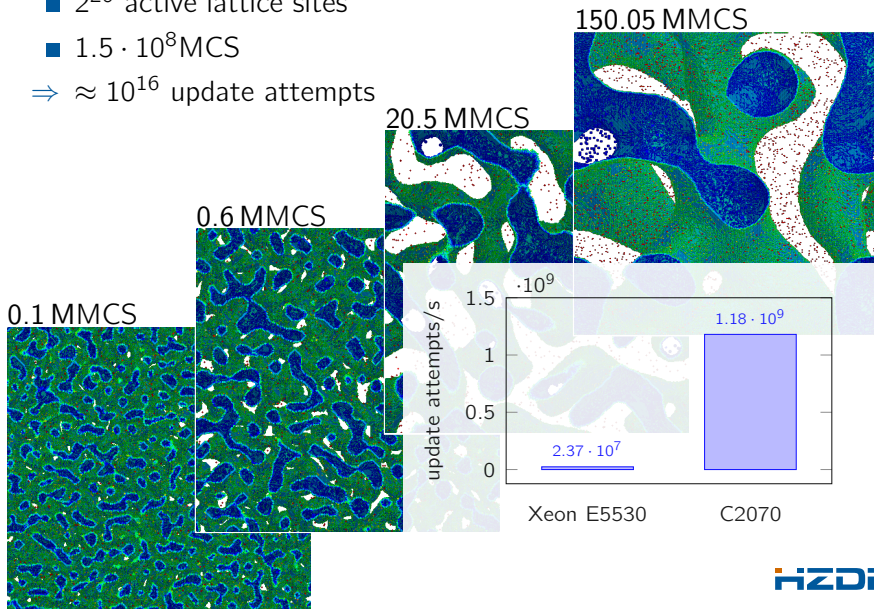
- simulated physical system evolves in discrete time steps Δt
- more accurate the larger the system
- updates are simple, but need to be applied in order
- uncorrelated updates—*We need those out-of-equilibrium.*

Why GPU and Massively Parallel Processing?

- 2^{26} active lattice sites

- $1.5 \cdot 10^8$ MCS

⇒ $\approx 10^{16}$ update attempts



Why GPU and Massively Parallel Processing?

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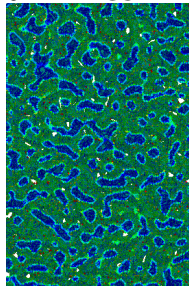
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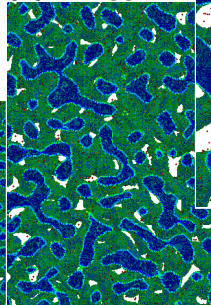
CPU ≈ 14 years

GPU ≈ 3.3 month

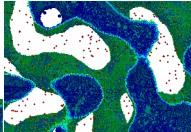
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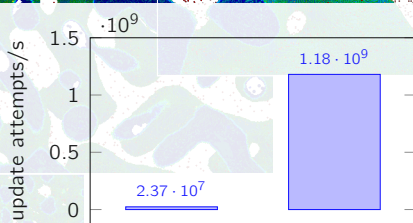
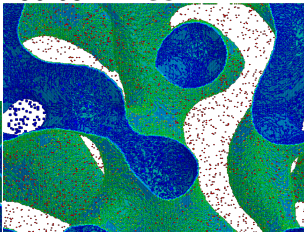
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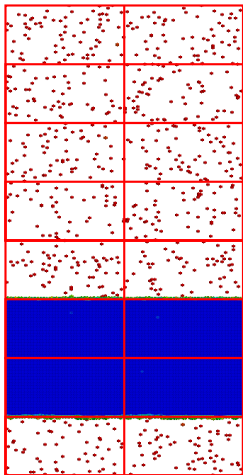
150.05 MMCS



Xeon E5530

C2070

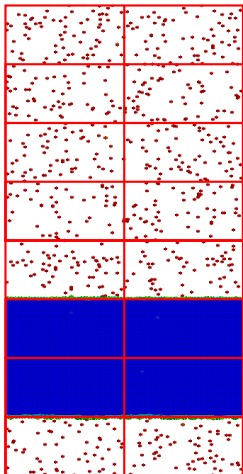
Broken Detailed Balance at Domain Boundaries



particle solubility in KMC

- device-layer domain boundaries break detailed balance:
 - move cannot be reversed at end of sweep
 - slower sweep at boundary

Broken Detailed Balance at Domain Boundaries



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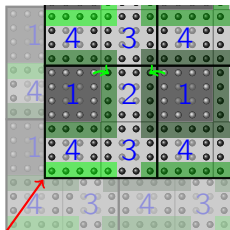
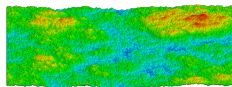
- device-layer domain boundaries break detailed balance:
 - move cannot be reversed at end of sweep
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$$c(\epsilon) = c_0 \cdot e^{-B\epsilon}$$

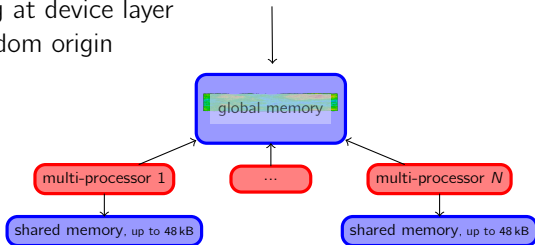
setup	c_0	B
aligned	1.08 ± 0.04	6.04 ± 0.03
random	1.00 ± 0.04	6.00 ± 0.03

⇒ solubility of particles at interface changed

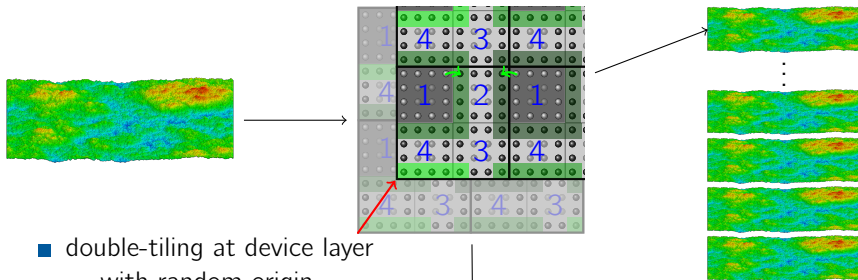
Multi-Surface Approach for GPUs



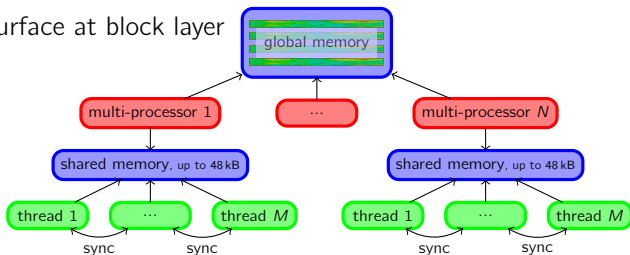
- double-tiling at device layer
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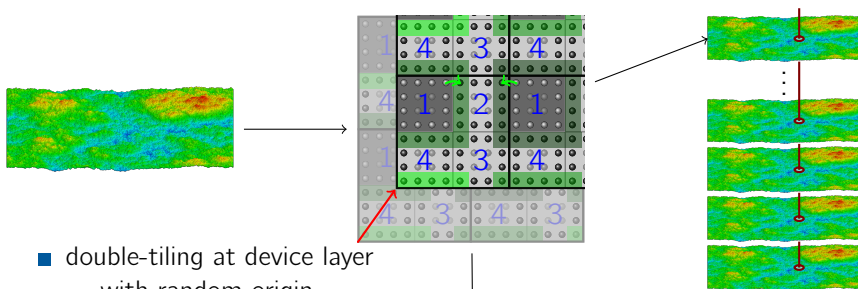
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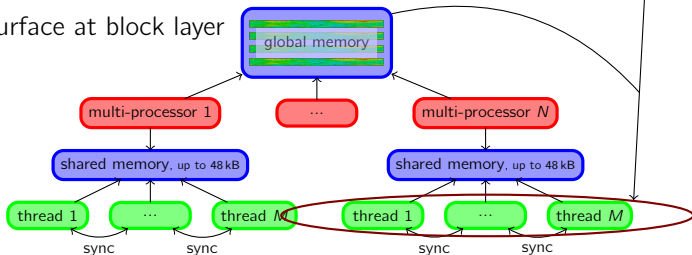
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Multi-Surface Approach for GPUs



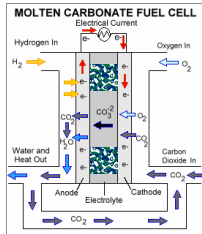
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Example I: Phase separation in inhomogeneous systems

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Nano Sponge as Porous Matrix Material



1

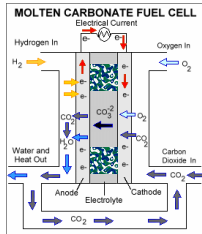
- porous matrix in Molten Carbonate Fuel Cells (MCFC)
- holding liquid electrolyte (Li_2CO_3) in nano-capillary matrix ($LiAlO_2$)



1MW MCFC2

^bhttp://www.fuelcell.no/fuel_cell_types_mcfc_eng.htm

Nano Sponge as Porous Matrix Material



1

- porous matrix in Molten Carbonate Fuel Cells (MCFC)
- holding liquid electrolyte (Li_2CO_3) in nano-capillary matrix ($LiAlO_2$)
- coarsening as degenerative process during operation—decreases capillarity

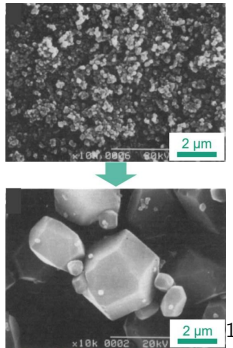


1MW MCFC2

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Coarsening in the Presence of Inhomogeneities

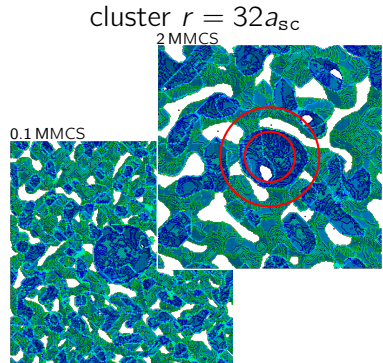
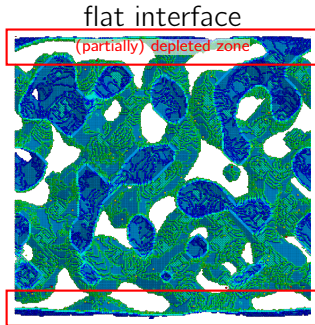
- large clusters with depleted zones have been observed in porous matrix at end of lifetime



¹Murai et al, *J. Electrochem. Soc.* **143** 2776 (1996)

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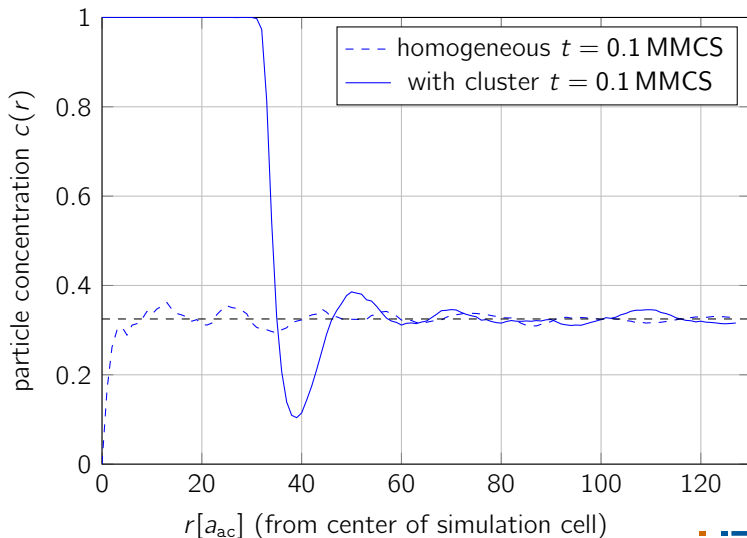
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$$\varepsilon = 2.0, c = 0.325, L = 2^8$$

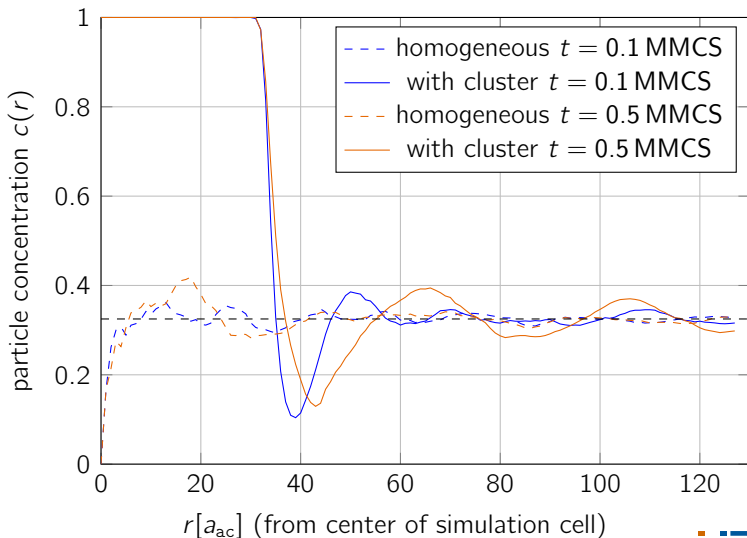
Sponge-Coarsening around Cluster

$$\varepsilon = 2.0, c = 0.325, L = 2^8$$



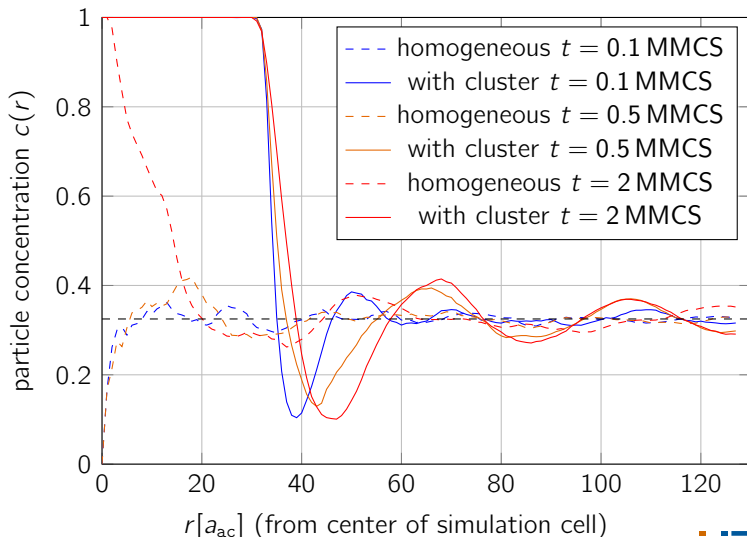
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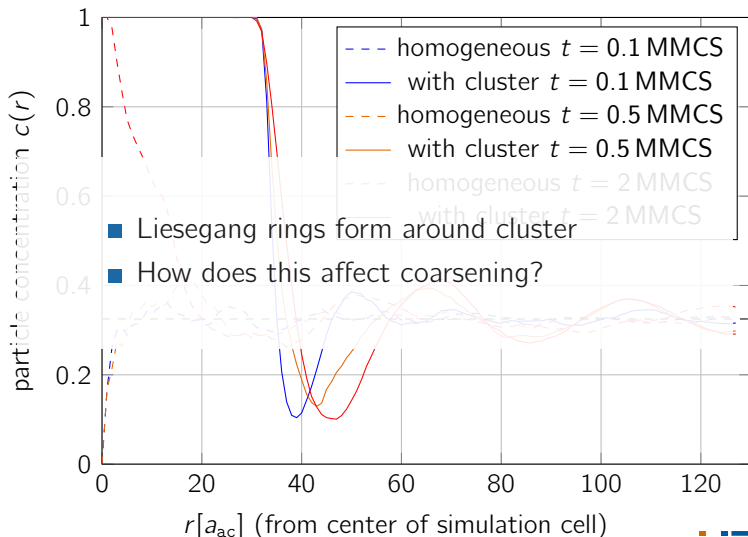
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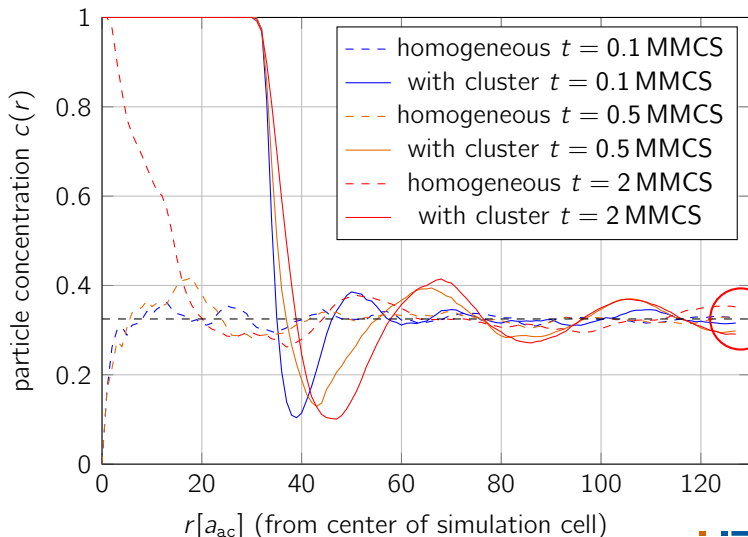
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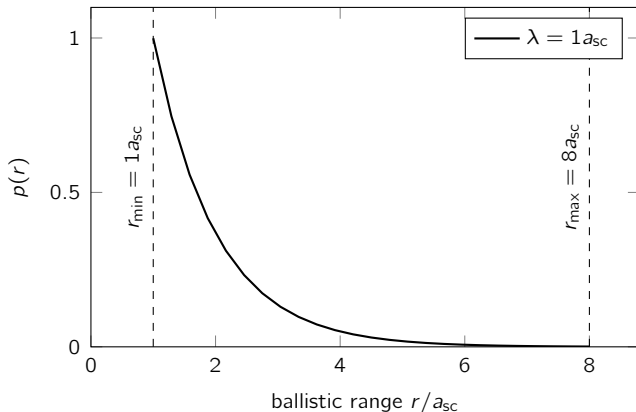
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Recoil Mixing: Model

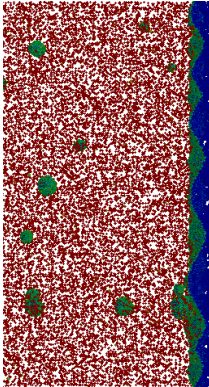
- incident ion hits lattice atoms creating a cascade of displacements
- energy gets divided among displaced atoms $\Rightarrow$ exponentially distributed range

$$p(r) \sim e^{-r/\lambda}$$

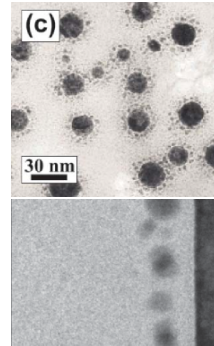


Inverse Ostwald Ripening

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Simulation of a (100)–interface
 $\varrho = 1 \times 10^{-3}$

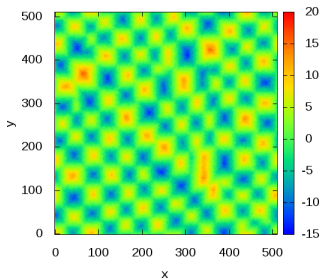


TEM pictures of Au nano clusters in SiO_2 matrix after irradiation with Au ions.

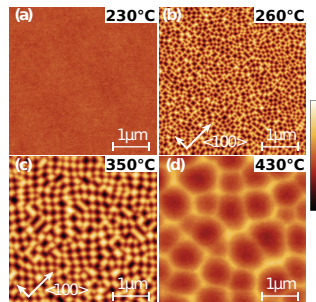
Heinig et al. Appl. Phys. A 77, 17 (2003)

Irradiation of Interfaces

Instability of buried Interfaces or Surfaces



Simulation of a (100)-interface
 $\varrho = 1 \times 10^{-3}$



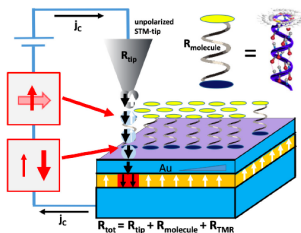
AFM picture of a Ge
(100)-surface after irradiation
with Ge ions.

Ou et al. (2013) submitted (arXiv 1303.5133)

Modeling in B2

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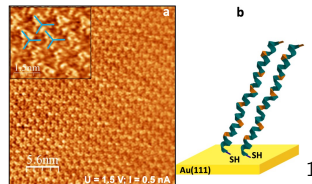
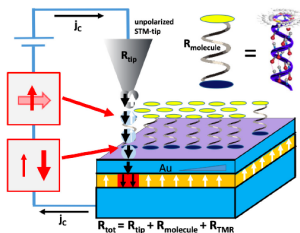
Modeling within SFB Subproject B2



¹Sharma, A., et al. *J. Mater. Chem. C* 8 11822 (2020)

Modeling within SFB Subproject B2

Modeling of structure formation of molecules using KMC



¹Sharma, A., et al. *J. Mater. Chem. C* 8 11822 (2020)

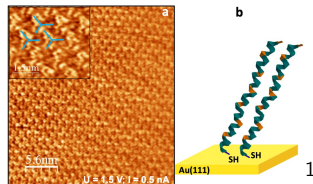
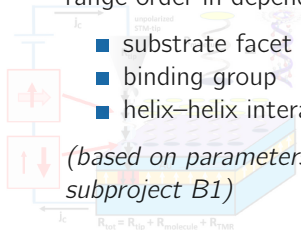
Modeling within SFB Subproject B2

Modeling of structure formation of molecules using KMC

1 modelling/prediction of long range order in dependence on:

- substrate facet
- binding group
- helix-helix interaction

(based on parameters from subproject B1)



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Modeling within SFB Subproject B2

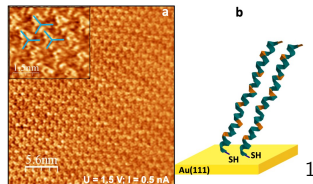
Modeling of structure formation of molecules using KMC

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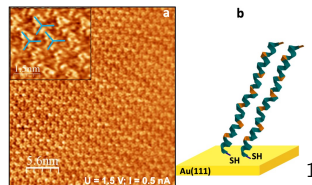
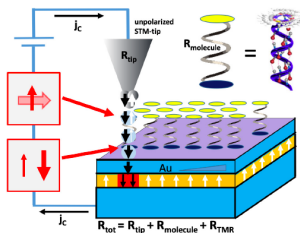
- 2 prediction of structure formation with multiple species



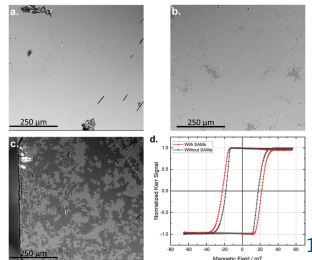
¹Sharma, A., et al. *J. Mater. Chem. C* 8 11822 (2020)

Modeling within SFB Subproject B2

Modeling of structure formation of molecules using KMC



Modeling of magnetic ordering of buried layer



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Modeling within SFB Subproject B2

Modeling of structure formation of molecules using KMC

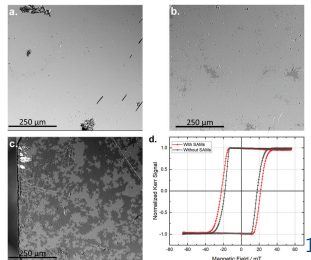
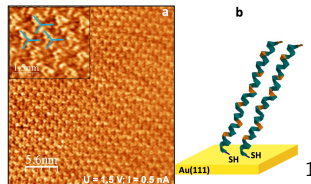
Modeling of magnetic ordering of buried layer

- 1 prediction of phase diagrams of magnetization, taking into account

- possibly periodic file from helical molecules

- disorder in molecular monolayer

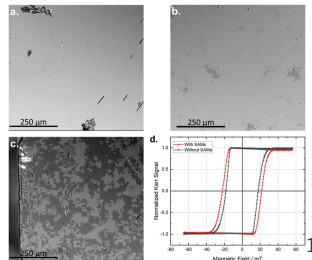
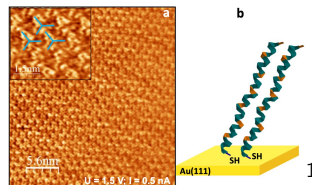
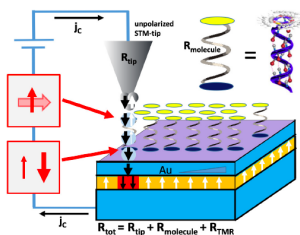
- 2 gauge magnetic interaction parameters against experimental measurements



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Modeling within SFB Subproject B2

Modeling of structure formation of molecules using KMC

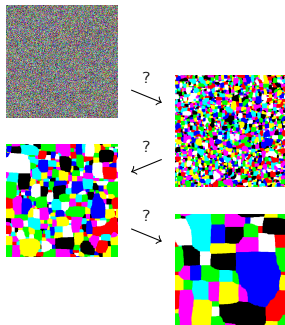


Modeling of magnetic ordering of buried layer

¹Sharma, A., et al. *J. Mater. Chem. C* 8 11822 (2020)

Dynamics vs Steady-State

if **kinetics** of interest
system out-of-equilibrium:

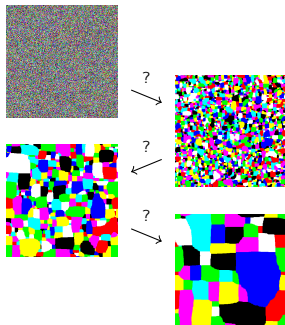


8-states Potts model, $\frac{J}{k_B T} = 5$

- optimal algorithm reproduces physical evolution

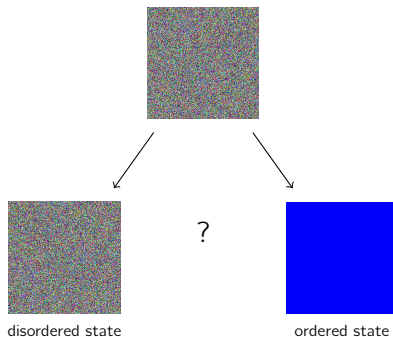
Dynamics vs Steady-State

if **kinetics** of interest
system out-of-equilibrium:



8-states Potts model, $\frac{J}{k_B T} = 5$

if only **final** state relevant
system (equilibrium) steady-state:



disordered state

ordered state

8-states Potts model

- optimal algorithm reproduces physical evolution

- optimal algorithm reaches equilibrium quickly

- Sibylle Gemming
 - Géza Ódor
 - Martin Weigel
 - Guido Juckeland
-
- Artur Erbe
 - Jörg Schuster
 - Peter Zahn
 - Karl-Heinz Heinig
 - Henrik Schulz
 - Nils Schmeißer
 - Michael Bussmann
-
- my other colleagues

Thank You.



J.Kelling@HZDR.de

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Selected Publications

- Kelling, J., Ódor, G.:
Extremely large-scale simulation of a Kardar-Parisi-Zhang model using graphics cards
Phys. Rev. E **84** 061150 (2011)
- Kelling, J., Ódor, G., Nagy, M. F., Schulz, H., Heinig, K.-H.:
Comparison of different parallel implementations of the 2+1-dimensional KPZ model and the 3-dimensional KMC model
Eur. Phys. J. ST **210** 175 (2012)
- Kelling, J., Ódor, G., Gemming, S.:
Bit-Vectorized GPU Implementation of a Stochastic Cellular Automaton Model for Surface Growth
IEEE International Conference on Intelligent Engineering Systems (2016)
- Kelling, J., Ódor, G., Gemming, S.:
Universality of 2+1 dimensional RSOS models
Phys. Rev. E **94** 022107 (2016)
- Kelling, J., Ódor, G., Gemming, S.:
Local scale-invariance of the 2 + 1 dimensional Kardar–Parisi–Zhang model
J. Phys. A **50** 12LT01 (2017)
- Kelling, J., Ódor, G., Gemming, S.:
Suppressing correlations in massively parallel simulations of lattice models
Comp. Phys. Commun. **220** 205 (2017)
- Ódor, G., Kelling, J.: Critical synchronization dynamics of the Kuramoto model on connectome and small world graphs, *Sci. Rep.* **9**, 19621, (2019)