Kinetic Monte Carlo method for the acceleration of defect evolution simulation

MAMMASMIAS Consortium

Anne Hemeryck, LAAS-CNRS, Toulouse, France Antoine Jay, LAAS-CNRS, Toulouse, France Nicolas Richard, CEA-DAM, Arpajon, France Layla Martin-Samos, CNR-IOM, Trieste, Italia Gabriela Herrero-Saboya, CNR-IOM, Trieste, Italia Miha Gunde, Institute Ruđer Bošković, Zagreb, Croatia

Nicolas Salles (CNR-IOM), MAMBA school

. . .

Kinetic Monte Carlo method for the acceleration of defect evolution simulation

Target:



Importance of the events catalog

Challenge of Off-Lattice kMC: Complexity increase

protons, ions



Silicon-based image sensor







> Molecular Dynamics: LAMMPS with Stilinger-Weber (SW) Potential









Molecular Dynamics VS kinetic Monte Carlo



kinetic Monte Carlo

Markovian chain of state $\{S_{i-1}, S_i, S_{i+1}\}$

- transition state or activated complex.



>Transition State Theory is an approach for modeling the rate of chemical reactions based on the idea of a

 \rightarrow The chemical reaction = a process where the reactants cross an energy barrier E_b to form products. This barrier is associated with a transition state where the reactants are transformed into products.

Lifetime of event as function T and activation barrier

1 <i>E</i>										J	
$\mathbf{L} \underline{\underline{L}_{ac}}$	Temp (K)	50	100	200	300	400	500	600	700	800	900
$t = -\rho^{k_B T}$	Energy (eV)										
	0.1	1.20E-003	1.10E-008	3.31E-011	4.79E-012	1.82E-012	1.02E-012	6.92E-013	5.25E-013	4.27E-013	3.63E-013
\mathcal{U}	0.2	1.44E+007	1.20E-003	1.10E-008	2.29E-010	3.31E-011	1.04E-011	4.79E-012	2.75E-012	1.82E-012	1.32E-012
t > 104	0.3	1.73E+017	1.32E+002	3.63E-006	1.10E-008	6.02E-010	1.06E-010	3.31E-011	1.45E-011	7.76E-012	4.79E-012
	0.4	2.08E+027	1.44E+007	1.20E-003	5.24E-007	1.10E-008	1.08E-009	2.29E-010	7.58E-011	3.31E-011	1.74E-011
	0.5	2.50E+037	1.58E+012	3.98E-001	2.51E-005	1.99E-007	1.10E-008	1.58E-009	3.98E-010	1.41E-010	6.31E-011
	0.6	3.00E+047	1.73E+017	1.32E+002	1.20E-003	3.63E-006	1.12E-007	1.10E-008	2.09E-009	6.02E-010	2.29E-010
	0.7	3.60E+057	1.90E+022	4.36E+004	5.75E-002	6.60E-005	1.14E-006	7.58E-008	1.10E-008	2.57E-009	8.31E-010
	0.8	4.33E+067	2.08E+027	1.44E+007	2.75E+000	1.20E-003	1.16E-005	5.24E-007	5.75E-008	1.10E-008	3.02E-009
	0.9	5.20E+077	2.28E+032	4.78E+009	1.32E+002	2.19E-002	1.18E-004	3.63E-006	3.02E-007	4.67E-008	1.10E-008
	1	6.25E+087	2.50E+037	1.58E+012	6.30E+003	3.98E-001	1.20E-003	2.51E-005	1.58E-006	1.99E-007	3.98E-008
	1.1	7.50E+097	2.74E+042	5.23E+014	3.01E+005	7.23E+000	1.22E-002	1.74E-004	8.31E-006	8.51E-007	1.44E-007
l > 10 J	1.2	9.01E+107	3.00E+047	1.73E+017	1.44E+007	1.32E+002	1.25E-001	1.20E-003	4.36E-005	3.63E-006	5.24E-007
	1.3	1.08E+118	3.29E+052	5.74E+019	6.90E+008	2.39E+003	1.27E+000	8.31E-003	2.29E-004	1.55E-005	1.90E-006
	1.4	1.30E+128	3.60E+057	1.90E+022	3.30E+010	4.36E+004	1.29E+001	5.75E-002	1.20E-003	6.60E-005	6.91E-006
nivacancies	1.5	1.56E+138	3.95E+062	6.29E+024	1.58E+012	7.93E+005	1.32E+002	3.98E-001	6.30E-003	2.82E-004	2.51E-005
	1.6	1.87E+148	4.33E+067	2.08E+027	7.57E+013	1.44E+007	1.34E+003	2.75E+000	3.31E-002	1.20E-003	9.11E-005
	1.7	2.25E+158	4.74E+072	6.89E+029	3.62E+015	2.62E+008	1.37E+004	1.90E+001	1.74E-001	5.12E-003	3.31E-004
	1.8	2.70E+168	5.20E+077	2.28E+032	1.73E+017	4.78E+009	1.39E+005	1.32E+002	9.11E-001	2.19E-002	1.20E-003
	1.9	3.25E+178	5.70E+082	7.55E+034	8.29E+018	8.69E+010	1.42E+006	9.11E+002	4.78E+000	9.32E-002	4.36E-003
	2	3.90E+188	6.25E+087	2.50E+037	3.97E+020	1.58E+012	1.44E+007	6.30E+003	2.51E+001	3.98E-001	1.58E-002
DFT -											

 $E_b = 0.84 \ eV$



$t > 10^{-4} s$



Monte Carlo (MC) method:

Random number to compute quantities

> Statistical method

kinetic Monte Carlo (kMC) method:

Random number to change system following a probability

> Probabilistic method



Transition State between discrete states over the time



> Frequentist representation

The time is divided in interval dt



some event happens during this time t



If we take *dt* small enough we can have, at most, one event by interval

we define the average rate:

$$r = \lim_{dt \to 0, t \to \infty} \sum_{t=1}^{n} \frac{N_{evt}}{t}$$

 $t = n \,.\, dt$

and the probability to have an event in time dt

$$p = r \,.\, dt$$

> Binomial representation

Binomial law b(n,p):

- n the number of experiment
- p (=r.dt) the probability to have a success

The probability to have x event in n trial:

P(X = x)

> Convergence to the Poisson law

We have $n \cdot p = n \cdot r \cdot dt = r \cdot t > 0$

The binomial law **b(n,r.dt)** converge to Poisson law of parameter $\lambda = r \cdot t$

P(X = x)

The probability of having x events in n attempt become -> the probability of having x event in time t

$$(n) = \binom{n}{x} (r \cdot dt)^{x} (1 - r \cdot dt)^{n-x}$$
 $\binom{n}{x} = \frac{n!}{x!(n-x)!}$

In the limit of small dt: $p \rightarrow 0$

$$n \to \infty$$

$$) = \frac{(rt)^{x}}{x!}e^{-rt}$$

Theorem:

- If N_t is number of event in time t and - \mathbf{T}_{n} is time between (n-1) and n-th event and follow the exponential law with same parameter K.t.

Poisson law of parameter λ

$$P(X = x) = \frac{(\lambda t)^{x}}{x!} e^{-\lambda t}$$

Probability of having x event in time t

Exponential law of parameter λ

$$P(X > t) = e^{-\lambda t}$$
 Probability of having

Then N_t is Poisson process of parameter K.t if and only if T_n are independent

ng an event after a time t

The time before the event happen: ln u



T_n independent means that the event happens at time t is independent of event happened in previous time

 $\mathbf{r}_{D?}$

Markov Chain hypothesis



Chain: $S_A \rightarrow S_B \rightarrow S_A \rightarrow S_C \rightarrow S_B \rightarrow S_D \rightarrow ...$ $T_1 T_2 T_3 T_4 T_5$

$$r_{A} = \sum_{I} r_{AI} \qquad T_{1} = -\frac{\ln u}{r_{A}}$$
$$r_{B} = \sum_{I} r_{BI} \qquad T_{2} = -\frac{\ln u}{r_{B}}$$

• • •

•
$$t = \sum T_i$$

T_n **independent** means that the event happens at time t is independent of event happened in previous time



Chain: $S_A \rightarrow S_B \rightarrow S_A \rightarrow S_C \rightarrow S_B \rightarrow S_D \rightarrow \dots$ \mathbf{T}_1 \mathbf{T}_2

$$\begin{split} \text{Master equation} \\ \frac{\partial P(\sigma, t)}{\partial t} &= \sum_{\sigma'} r_{\sigma' \to \sigma} \,. \, P(\sigma', t) - r_{\sigma \to \sigma'} \,. \, P(\sigma, t) \end{split}$$



kinetic Monte Carlo (kMC): Resume





?

kinetic Monte Carlo (kMC): Resume



1. Compare the system to the catalog of event 2. Give the list of possible event to the kMC algorithm 3. Return the site and the event selected

4. Apply the event on the selected site



Kinetic Monte Carlo algorithm

Residence time algorithm (Bortz, Kalos and Lebowitz)

Step 0 - Set the time t = 0, initial configuration

Step 1 - Form a list of all the rates r_{ii} of all possible transitions P_i in the system

Step 2 - Calculate the partition function $Z = \sum r_{ij}$ for i = 1,..., N where N is the total number of transitions.

Step 3 - Get a uniform random number $n \in [0, 1]$ Step 4 - Find the event to carry out *i* by finding the i for which $P(ik) < n \cdot Z < P(im)$

Step 5 - Apply event *i*, change the local atomic configuration Step 6 - Find all P_i and recalculate all r_{ii} which may have changed due to the transition Step 7 - Get a new uniform random number $n \in [0, 1]$ Step 8 - Update the time with t = t + dt where: Step 9 - Return to step 1

Rate of the event Probability

Stochastic choice of event

Temporal evolution







First reaction algorithm (Gillepsie)

Step 0 - Set the time t = 0, initial configuration

Step 1 - Form a list of all the rates r_{ij} of all possible transitions P_i in the system

Step 3 - For each possible event $S_i \rightarrow S_j$, get a random number $u \in [0, 1]$ and compute associated time $\tau = -\frac{\ln u}{-1}$

Step 4 - Select the event $S_i \rightarrow S_j$ with the lowest τ

Step 5 - Apply selected transition, and increase time by τ . Step 6 - Return to step 1 (advantage: previous computations are stored, only nearest neighbors and associated events are updated)

 r_{ii}



Rate of the event Probability

Stochastic choice of event

> Temporal evolution







Standard algorithm / Rejection

Step 0 - Set the time t = 0, initial configuration

Step 1 - Form a list of all the rates r_{ii} of all possible transitions P_i in the system

Step 2 - Calculate the partition function $Z = \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{i$ number of transitions. i,j

Step 3 - Choose a transition at random $S_i \rightarrow S_j$ Step 4 - Get a random number $u_1 \in [0,1]$ Apply event if $u_1 < \frac{r_{ij}}{Z}$

Step 5 - If transition accepted, pick a random number $\mathcal{U}_2 \in [0,1]$, and increase time $\ln u_{\gamma}$ Otherwise the event is rejected

Step 6 - Return to step 1

$$r_{ij}$$
 for i = 1,..., N where N is the total

Rate of the event Probability

Stochastic choice of event

Temporal evolution

$$\epsilon = \frac{N_{rejected}}{N_{tot}}$$







kinetic Monte Carlo (kMC): Resume



kinetic Monte Carlo:

- Versatile method, large kind of system can be describes
- Large system dimension & long time evolution

Difficulties:

- Description depend on the list of event
- Not fully Parallelizable but some optimization exist

kinetic Monte Carlo (kMC): Resume





Approximation of the system by a grid of point



+ events catalog

OnLattice kMC application

- Methodology of industrial interest
- to optimize manufacturing processes and reduce costs
- to simulate processes and minimize defects

Synopsys® TCAD

erest ocesses and reduce costs inimize defects



EXPERIMENTAL EVIDENCES



TEM as-implanted

- Amorphous layer in RT implant
- No amorphous layer for 150°C and 500°C implants
- > Damaged area in 150 °C case

EXPERIMENTAL EVIDENCES



TEM as-implanted

- Amorphous layer in RT implant
- No amorphous layer for 150°C and 500°C implants
- Damaged area in 150 °C case

TEM post-annealing

- Defect location: Amorphous layer has an impact on the depth of the defects
- Defect type: Dislocation loops (DLs) in RT and 150°C / {311} defects in 500°C case
- > Defect density: More interstitials trapped in RT and 150°C than in the 500°C

l in	l in {311}		
RT	150°C	500 °C	
1.4 10 ¹⁴ cm ⁻²	1.6 10 ¹⁴ cm ⁻²	5.4 10 ¹² cm	



Modeling – KMC default calibration





KMC post-annealing Defect type:

- RT: DLs simulated agree with TEM
- > 150°C: DLs simulated agree with TEM
- > 500°C: DLs simulated different from
 {311} in TEM

Defect density

	RT	150°C	
TEM	1.4 10 ¹⁴ cm ⁻²	1.6 10 ¹⁴ cm ⁻²	5.1 10 ¹²
KMC	1.4 10 ¹⁴ cm ⁻²	2.4 10 ¹⁴ cm ⁻²	1.9 10 ¹⁴
	KMC simulation for RT and but not for \$	ons are accurate 150°C cases 500°C implants	



Modeling – KMC default calibration





What parameter to change?

- Same annealing sequence after the implants
 → The difference should be observed in the as-implanted state of the material
- What are the defects produced by a single As atom implanted?

In 500°C SMIC with size < 7

Calibration of small interstitial clusters (SMICs) is required

Histogram of interstitials-vacancies defects for an As atom implanted at RT, 150°C and 500°C

Modeling – KMC – New calibration from



What are the SMIC energies in literature ?

• Two SMIC types in literature :

Chain-like and Araï like (001-Compact)

Ref. MD simulation - Marques (2019). Acta Mater. 166, 192-201





SMIC of Araï type should be considered in detail

Modeling – KMC – New calibration from



What if the MD trend for Araï SMICs energies is used to fit experiments?



Defect type accurate

TE

KM

Reproduce the extended defects trends for the 3 implantations

Need for fine atomistic ingredient in TCAD







OnLattice kMC: Resume

> List of events

Known in advance: KMC only uses events Assumption that events are well defined and that their transition rates are constant or can be calculated

However, in complex systems, events may be dependent on the environment and local conditions, making them difficult to describe. Same for long range interaction.

Limits:

Calculation of Rates: The accuracy of KMC is highly dependent on the *accuracy of transition rates*. Calculating these rates accurately can be difficult, especially for complex systems.

Modelling error: *Incorrect transition rates* can lead to <u>bad results</u>, reducing the reliability of simulations.

Model accuracy: The importance of *accuracy in transition rate* models for reliable results.



> On the fly kinetic Monte Carlo / Adaptive KMC (AKMC)



Approximation of the system by E/F Engine



+ events catalog

> On the fly kinetic Monte Carlo / Adaptive KMC (AKMC)



The **representation of the event** has to take into account the degrees of freedom of the E/F Engine





> On the fly kinetic Monte Carlo / Adaptive KMC (AKMC)



Simulated size, time



Simulated size, time

OpenSource OffLattice kMC

Adaptative events catalog needs an automatic way to search new events

ARTn (Activation Relaxation Technique nouveau) Normand Mousseau
plugin-ARTn (https://gitlab.com/mammasmias/artn-plugin)
Software: kART (https://normandmousseau.com/fr/research/kinetic-art)

Dimere method (Graeme Henkelman and Hannes Jonsón)

Software: eON (https://github.com/TheochemUI/eOn)



Accurate Transition states Exhaustive list of events









> Output of Molecular Dynamics + kinetic ART (Code kMC)



After each kMC step (event application) the system is relaxed by the E/F Engine





Antoine Jay (LAAS-CNRS)





Antoine Jay (LAAS-CNRS)

Long time simulation: Feedback





> Output of kART -> ab initio calculation

The defect structure of complex defect are really different than those expected ab initio words

Gabriela Herrero-Saboya: Talk tomorrow



Simulated size, time



Long time simulation: Feedback





> Output of kART -> ab initio calculation

The defect structure of complex defect are really different than those expected ab initio words



Gabriela Herrero-Saboya: Talk tomorrow

OffLattice kMC (kMC+ARTn+E/F Engine)

- Deal with complex system on long time
- Computation cost of relaxation/research event

How to keep/reuse the information already computed?

Event configuration in continuous configuration space





Thank You for your attention