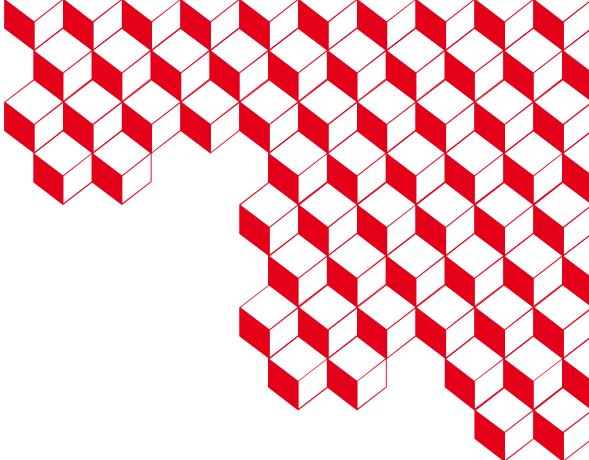


May 26th-28th, 2025 - Vienna, Austria

7th International Workshop on Models and Data for Plasma-Material
Interactions in Fusion Devices (MoD-PMI)



Multiscale modeling of diffusion phenomena in nuclear fuels: from the atomic scale to phase-field polycrystalline microstructure evolution simulations

Luca MESSINA¹,

Tommaso BARANI¹, Clément INTROIINI¹, Petra OSPITAL¹, Maciej J. KARCZ¹,
Thomas SCHULER², Eiji KAWASAKI³, Frédéric SOISSON², Marjorie Bertolus¹,
Emeric BOURASSEAU¹

¹ CEA IRESNE, DEC, Cadarache, France

² CEA ISAS & Université Paris-Saclay, France

³ CEA LIST & Université Paris-Saclay, France

luca.messina@cea.fr

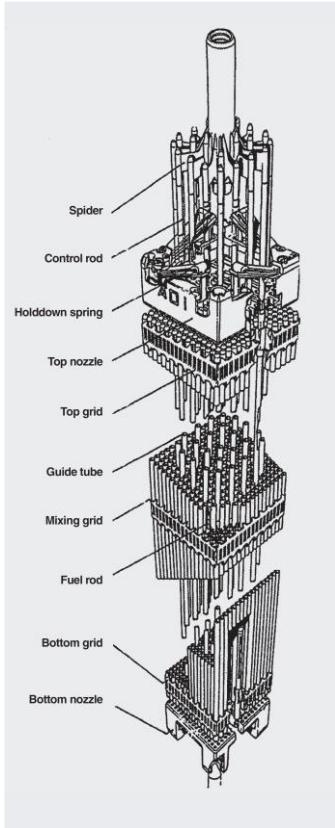


Outline

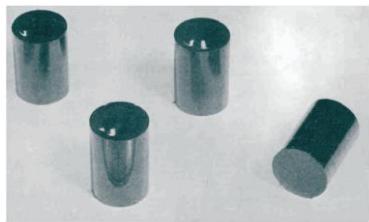
- 1. Nuclear fuels and fission gases**
- 2. Phase-field modeling of gas and microstructure evolution**
- 3. Atomic-scale diffusion coefficients**
- 4. AI approach to the partition function of disordered compounds**
- 5. Conclusions and possible applications to fusion materials**



Nuclear fuels

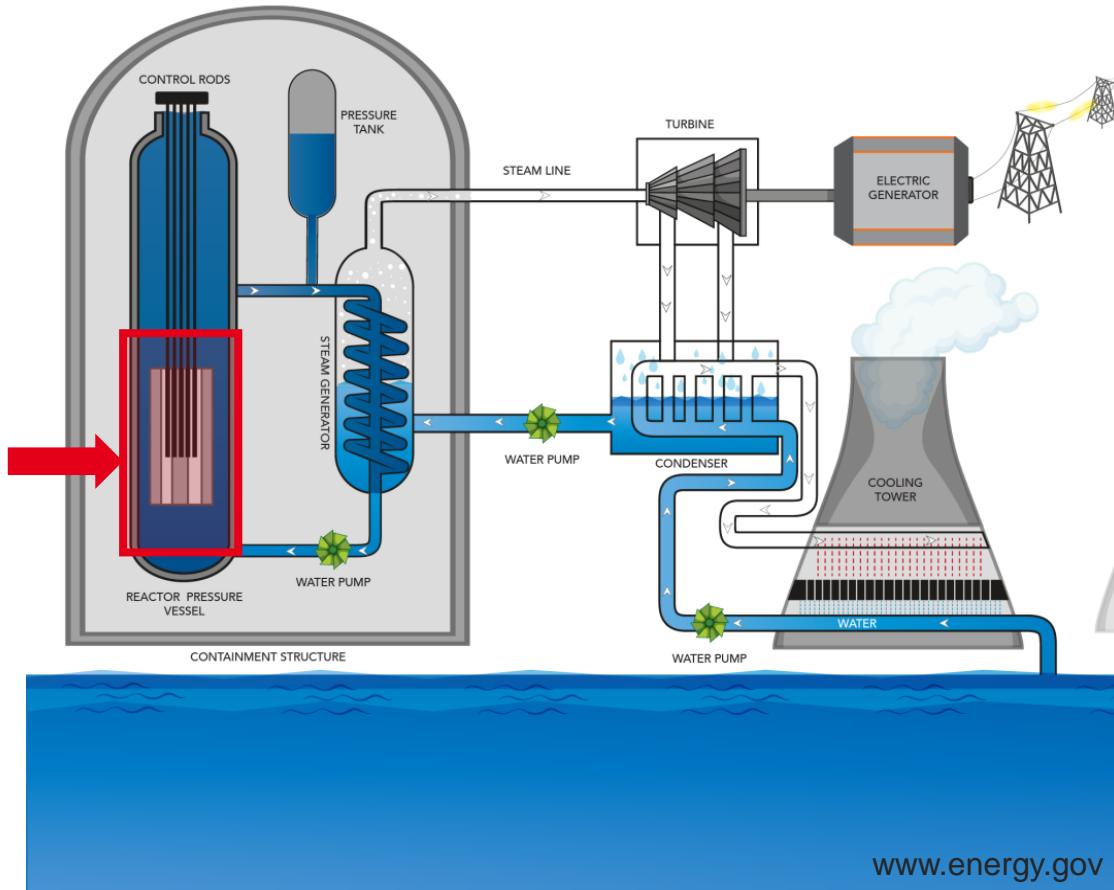


PWR Fuel assembly^[1]



Fuel pellets^[1]

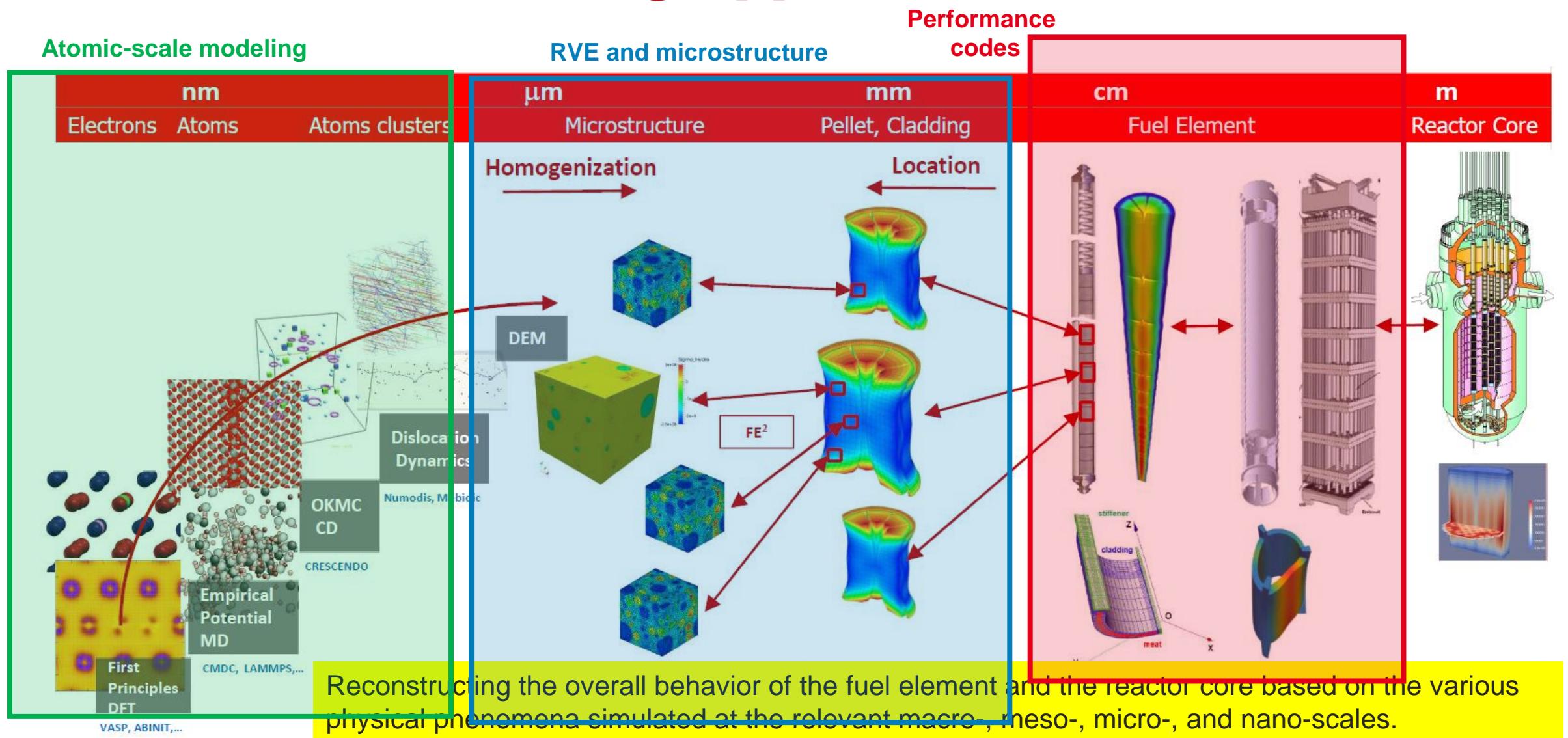
- Development of **nuclear fuels**: safety, efficiency, resources optimization, waste minimization
 - Complex behaviour of fuel and fission products, with coupled **multi-physics** and **multiscale phenomena**
 - Operational margins mainly determined by **thermomechanical behavior** of fuel elements
 - Consequences of **accidents** depend on physico-chemical fuel behaviour (potential release of radioactive material)
- ⇒ **Important to understand and be able to predict the fuel behavior under irradiation**



[1] Nuclear Fuels, CEA, ISBN 978-2-281-11345-7 (2008)



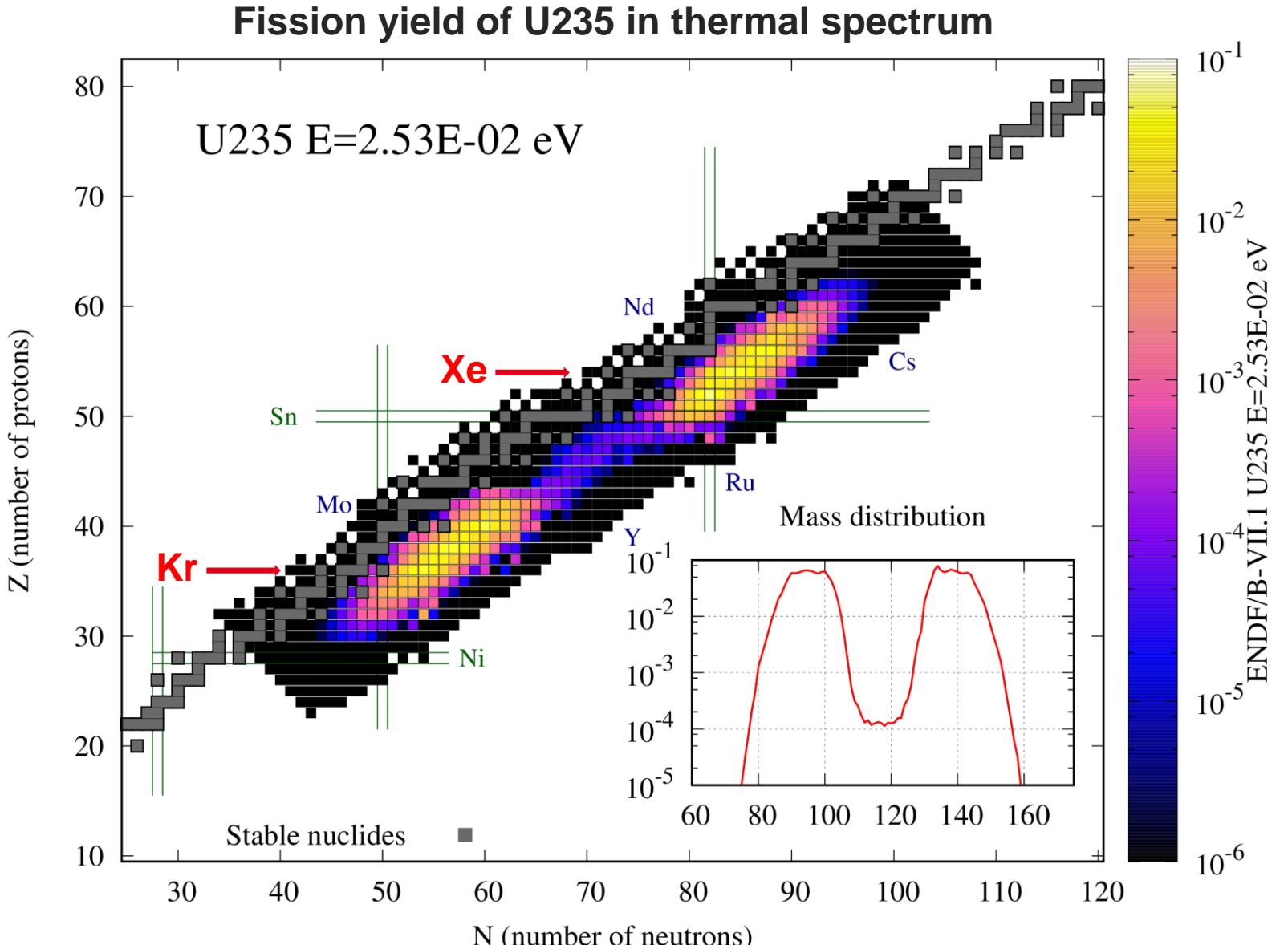
Multiscale modeling approach to nuclear fuels





Fission gases

- Accumulation of inert gases during irradiation
 - Xenon (about 20%)
 - Krypton (about 10%)
 - Helium (by (n, α) reactions and α decay)
- Relevant impact on fuel behavior



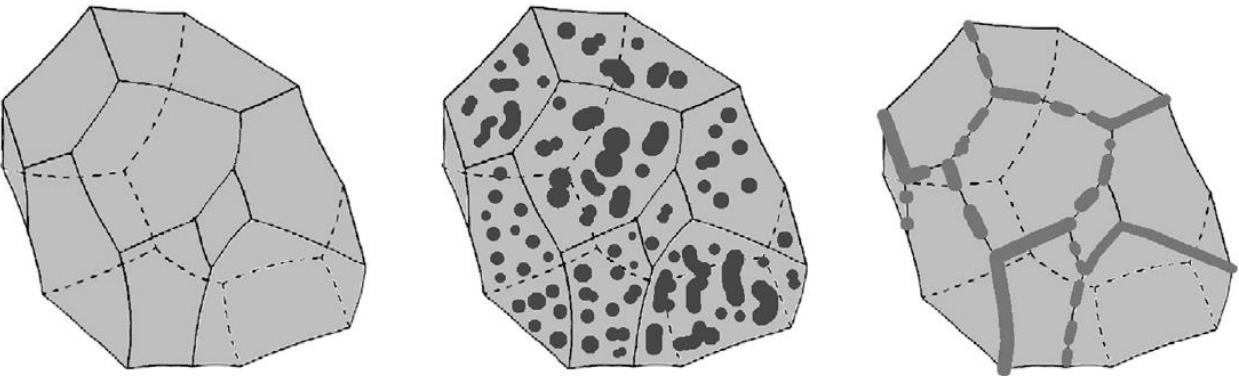
Fission gas phenomenology

- Fission gas production (Xe, Kr, He) => atomic-scale intra-granular diffusion
- Intra-granular bubble formation and coalescence
- Gas accumulation on grain boundaries => inter-granular bubbles
- Growth and interconnection of inter-granular bubbles => **gas release** in fuel rod free volumes

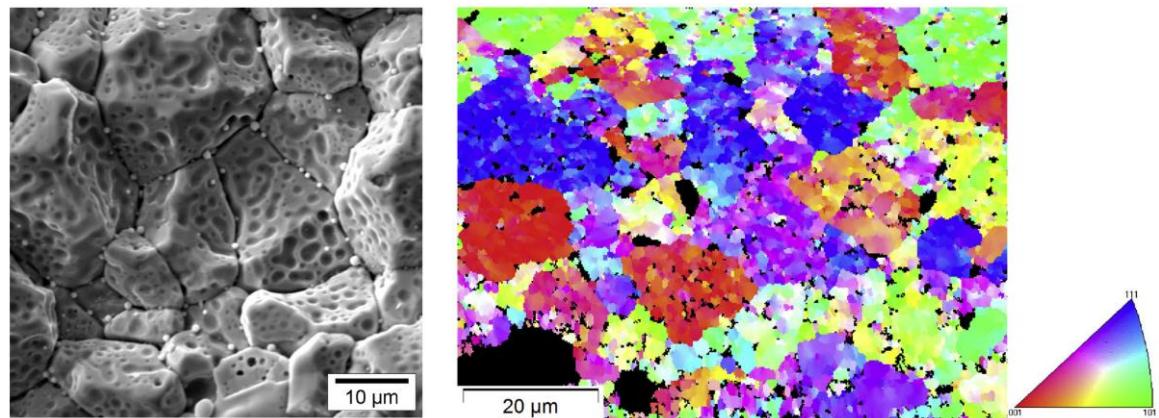
- Fission gas evolution ⇔ microstructure evolution
- Fission gas evolution ⇔ fuel pellet **thermomechanical behavior**

- Deformation, **swelling**
- Restructuring phenomena
 - high-burnup structures (periphery)
 - **formation of sub-grains** (central area)

M. Tonks et al., J. Nucl. Mater. 504 (2018), 300



J. Noirot et al., Nucl. Eng. Tech. 50 (2018), 259



Sub-grain formation in the central area with gas bubble network

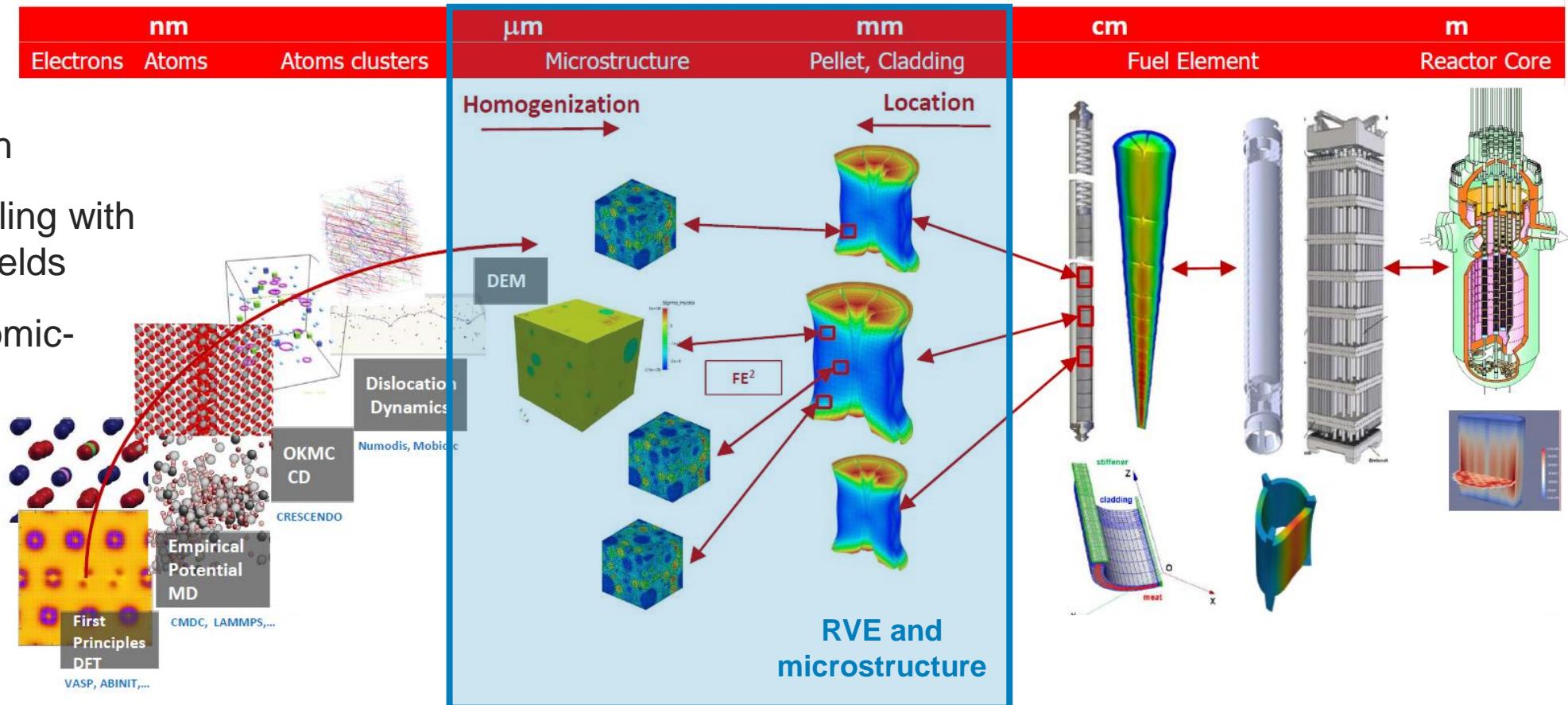
FIB/SEM electron backscattered diffraction (EBSD) inverse pole figure map of the polished surface



Objective

Development of 3D simulation tool for fission gas behavior in polycrystalline microstructures, integrated with evolution of thermomechanical fields

- Phase field approach
- Self-consistent coupling with thermomechanical fields
- Parameters from atomic-scale modeling





Outline

1. Nuclear fuels and fission gases
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Phase-field model of gases in polycrystalline microstructures

- 2 phases (matrix/bubbles), with p grains and n bubbles
- One order parameter η_1, \dots, η_p for each matrix grain, one unique order parameter η_b for bubbles
- Chemical potential (\Leftrightarrow concentrations) fields of U-vacancies c_v and substitutional gas atoms c_g
- Diffuse interfaces (continuous variation of each field from 0 to 1)

$$\Omega = \int_V \left[f_{\text{chem}} + f_{\text{interface}} + \frac{\kappa}{2} |\nabla \eta|^2 (+f_{\text{el}} + \dots) \right]$$

**ENERGY
MINIMIZATION**

- Evolution of order parameters : Allen-Cahn equation

$$\frac{\partial \eta_{\alpha i}}{\partial t} = -L(\mathbf{r}) \left[m \left(\eta_{\alpha i}^3(\mathbf{r}) - \eta_{\alpha i}(\mathbf{r}) + 2\eta_{\alpha i}(\mathbf{r}) \sum_{\beta=1}^N \sum_{j=1, \alpha i \neq \beta j}^{p_\beta} \gamma_{\alpha i \beta j} \eta_{\beta j}^2(\mathbf{r}) \right) - \kappa \nabla^2 \eta_{\alpha i}(\mathbf{r}) + \sum_{\beta=1}^N \frac{\partial h_\beta}{\partial \eta_{\alpha i}} \omega_\beta(\mathbf{r}) \right]$$

- Evolution of chemical potentials: Cahn-Hilliard equation

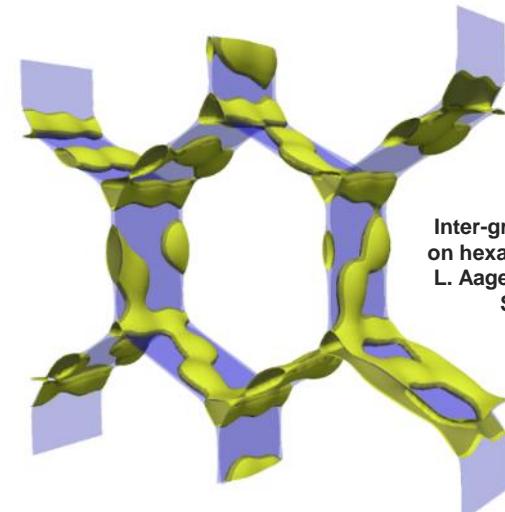
$$\frac{\partial \mu_I(\mathbf{r})}{\partial t} = \frac{1}{X_{II}(\mathbf{r})} \left(\nabla \cdot M_{II}(\mathbf{r}) \nabla \mu_I(\mathbf{r}) - \sum_{\beta=1}^N \sum_{i=1}^{p_\beta} \frac{\partial \rho_I(\mathbf{r})}{\partial \eta_{\beta i}} \frac{\partial \eta_{\beta i}(\mathbf{r})}{\partial t} + s_I(\mathbf{r}) \right)$$

Input parameters

- | | |
|---|---|
| <ul style="list-style-type: none"> • Solid/gas phase free energies • Gas/defect formation energies • Gas/defect diffusion coefficients | <ul style="list-style-type: none"> • Gas/defect source term • Grain boundary properties • Restructuring onset criteria |
|---|---|

Physics ingredients

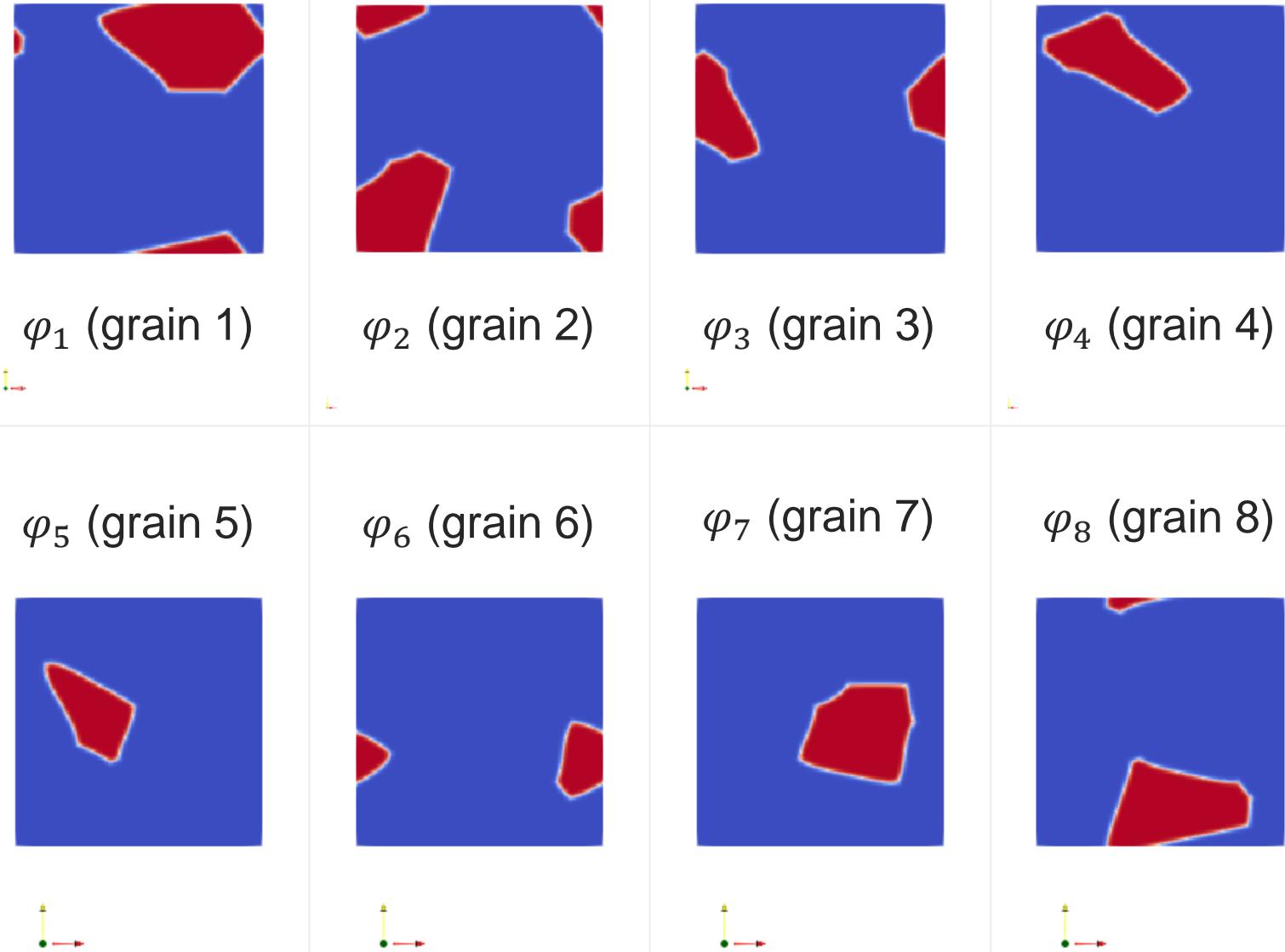
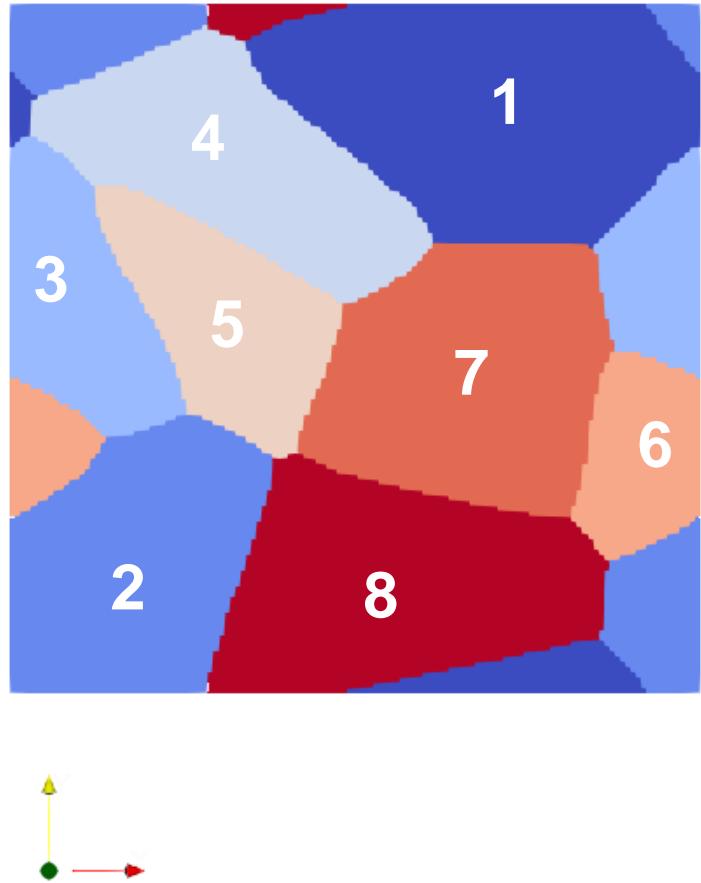
- Grain interface/surface energy \Rightarrow evolution of grain/bubble shapes
- Gas/defect irradiation source terms
- Gas/defect diffusion towards grain boundaries (effective diffusion neglecting intragranular heterogeneities)
- Phase free energies and non-equilibrium conditions



Inter-granular bubble evolution on hexagonal regular polycrystal
L. Aagesen et al., Comp. Mater. Sci. 161 (2019), 35

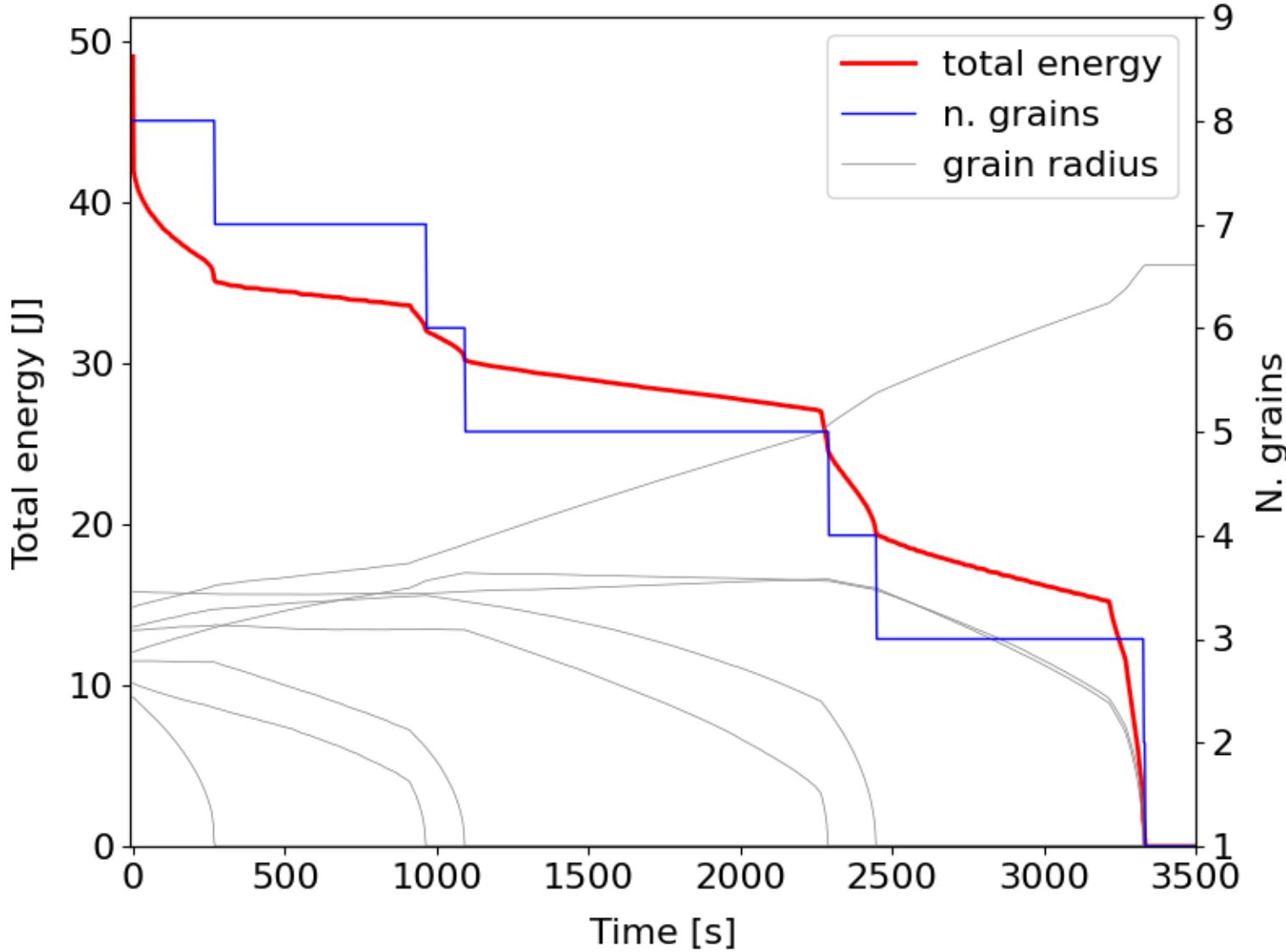
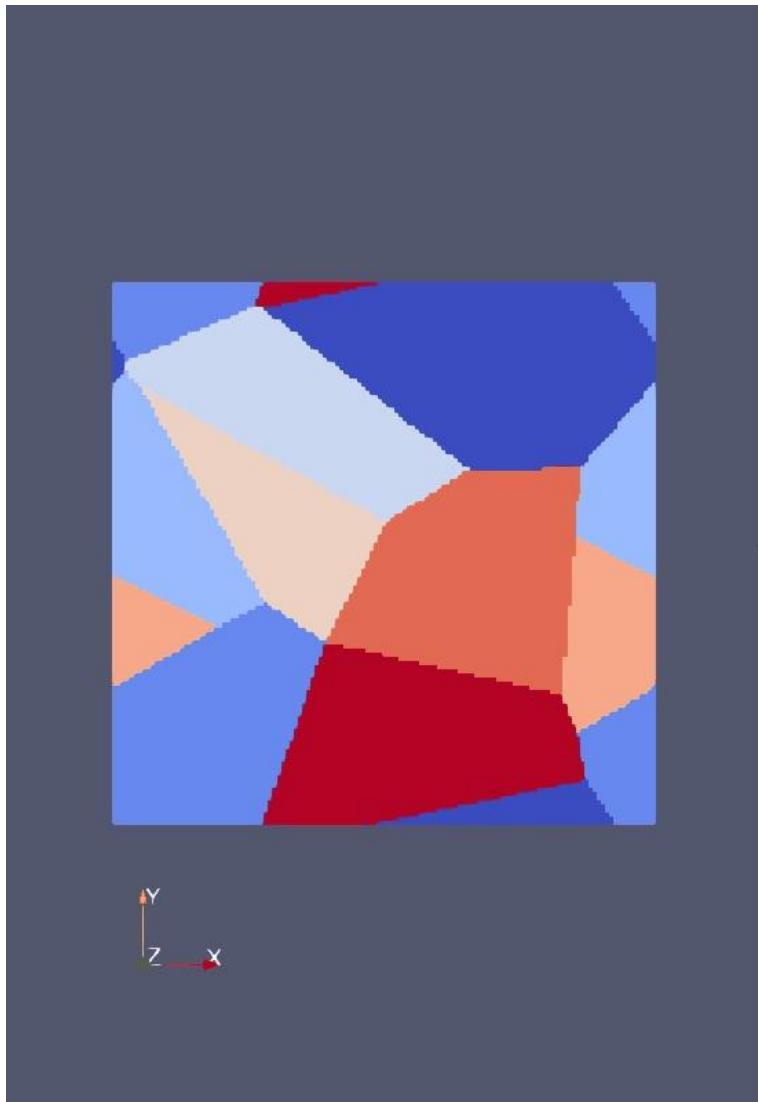


Polycrystal representation in phase field



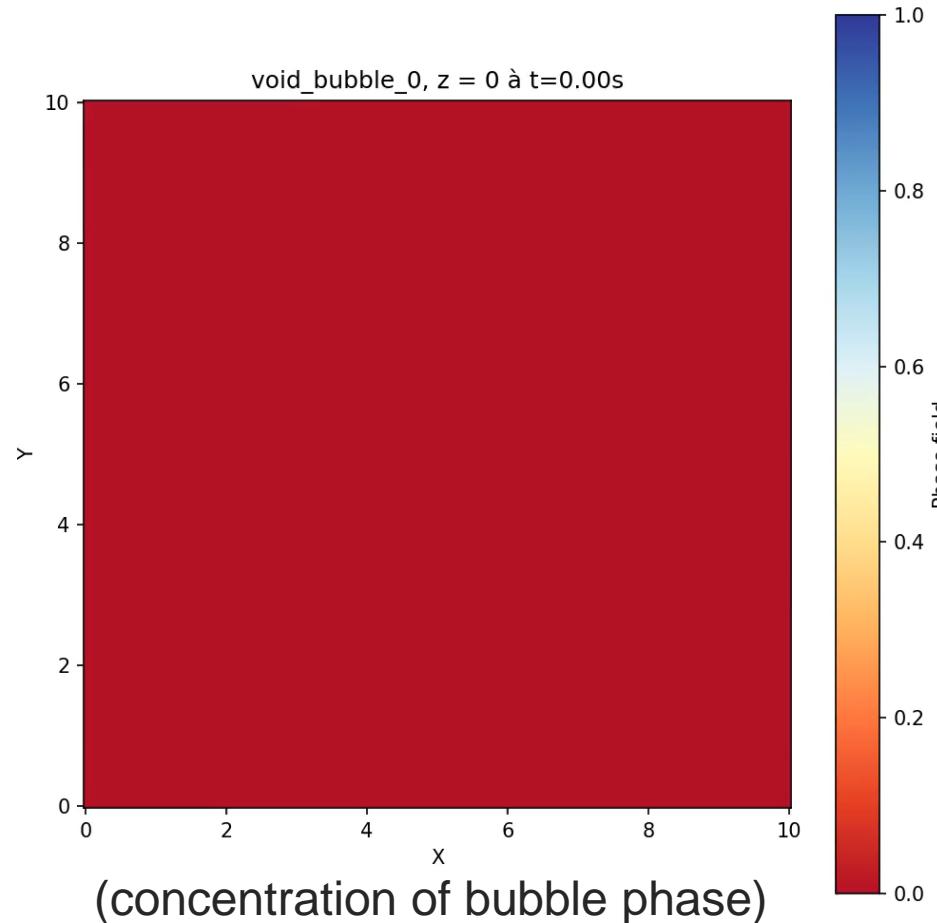


Example - Grain growth (2D)



Example – Intragranular bubble nucleation

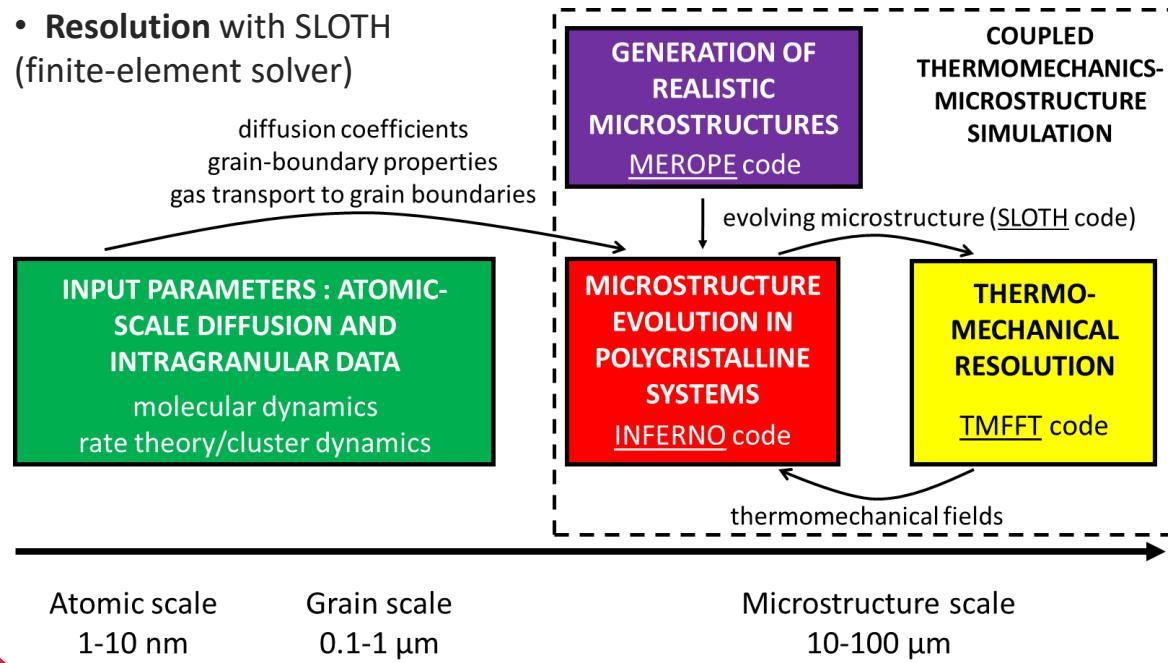
Non-homogeneous random fission source
(generation of gas and point defects)



The INFERNO code

Coupled microstructure-mechanics resolution

- Phase-field guided microstructure evolution, receiving at each time step information about mechanical fields from FFT thermomechanics codes.
- Initial microstructure generated by MEROPE from experimentally observed microstructures.
- Resolution with SLOTH (finite-element solver)



Dedicated computational platform to solve custom multiphase-field models

- Fission gases
- Sintering
- Porosity evolution
- ...

$$\Omega = \int_V \left[f_{\text{chem}} + f_{\text{interface}} + \frac{\kappa}{2} |\nabla \eta|^2 + f_{\text{elastic}} + f_{\text{viscoplastic}} \right]$$

provided at each time step by independent solver of mechanics equations

Code coupling

MEROPE

Microstructure generator for initial conditions of phase field simulations

TMFFT

Fast-Fourier transform solver for thermo-mechanics equations

SLOTH

3D HPC finite-element phase-field solver based on the MFEM* library

*mfem.org

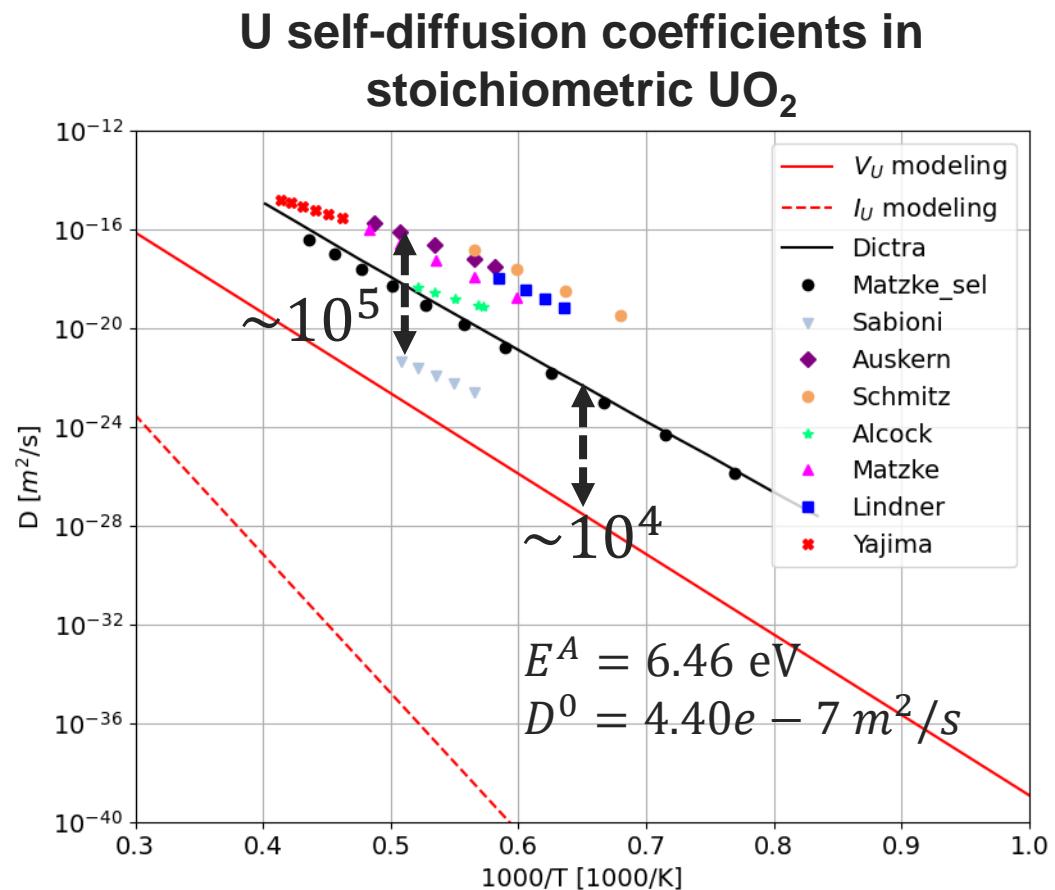


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Cationic self-diffusion in UO_2

- Dispersion among experimental values
 - Mainly explained by difficulty to control stoichiometry
- Large gap between modeling (lines) and experimental values (points) for isolated defects
- Gap could be explained by:
 - Thermal-expansion effect on migration energies [1]
 - Contribution of vacancy clusters



[1] W. D. Neilson *et al.*, J. Phys. Chem C **128**, 21559 (2024)

Moore, J. Solid State Chem. 203 (2013)



How to obtain a diffusion coefficient?

- Experimentally:
 - Direct measurements (ex. tracers)
 - Indirect measurements (ex. from gas release experiments during thermal ramps)
- Based on modeling:
 - Mean square displacement in Molecular Dynamics (MD) or Kinetic Monte Carlo (KMC) simulations
 - Analytical methods (KineCluE [1] and other methods)
 - MD => needs interatomic forces based on interatomic potentials or ab initio MD
 - KMC / analytical methods => need atomic-scale input such as:
 - Activation energies (defect formation + migration)
 - Prefactors

$$D = g \nu \exp\left(-\frac{G^{\text{act}}}{k_B T}\right) = g \underbrace{\exp\left(-\frac{G^{\text{for}}}{k_B T}\right)}_{\text{DEFECT CONCENTRATION}} \nu \exp\left(-\frac{G^{\text{mig}}}{k_B T}\right) \underbrace{\nu}_{\text{JUMP FREQUENCY}}$$

Defect activation free energy Formation free energy Migration free energy

Attempt frequency (prefactor)

[1] T. Schuler *et al.*, Comput. Mater. Sci. 172 (2020)

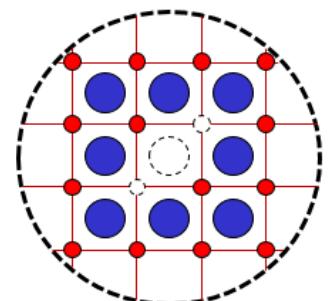
Self-diffusion by small vacancy clusters

- **Dilute limit hypothesis :** The material defect clusters are low enough in concentrations to consider them non interacting at least until they achieve thermodynamic equilibrium
- The diffusion coefficient can be expressed with KineCluE [1] (KINetic CLUster Expansion) as:

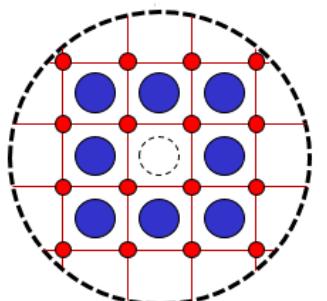
$$D_{U^*} = \sum_c P_c \frac{L_{U^*U^*}(c)}{c_U^*} ; \quad P_c = \frac{[c]}{\sum_{c'} [c']}$$

- We need to determine for each cluster :
 - 1) The probability P_c for the U tracer to be in cluster c : *Which defect cluster is the most probable ?*
 - 2) The cluster transport coefficient $D_{U^*}(c)$: *How fast do the defect clusters transport the U tracer atoms*

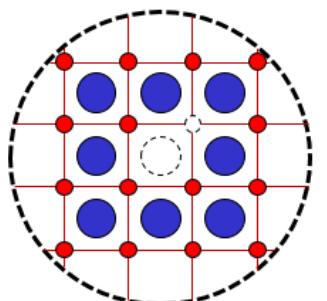
Studied defect clusters :
(and their multiple internal configurations)



Schottky defect
(SD or $V_U V_O V_O$)

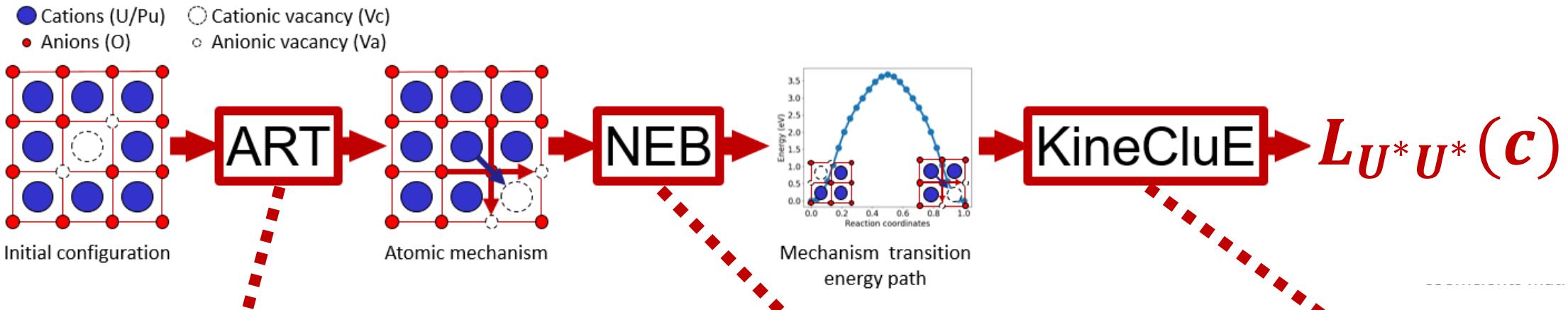


UO divacancy
($V_U V_O$)



U vacancy
(V_U)

Cationic self-diffusion coefficients with combined ART + KineCluE approach



ART (Activation Relaxation Technique)

- Searches energetic transitions by random explorations of the potential energy surface

Mousseau, N. J. Phys. B 2012 (2012)

NEB (Nudged Elastic Band)

- Breaks down energy transitions into migration mechanisms
- Computes the migration barriers of the mechanisms based on the initial and final atomic configurations

Jonsson, H. Classical and Quantum Dynamics in Condensed Phase Simulations p.385-404 (1998)

KineCluE (Kinetic Cluster Expansion)

- Computes the cluster transport coefficient

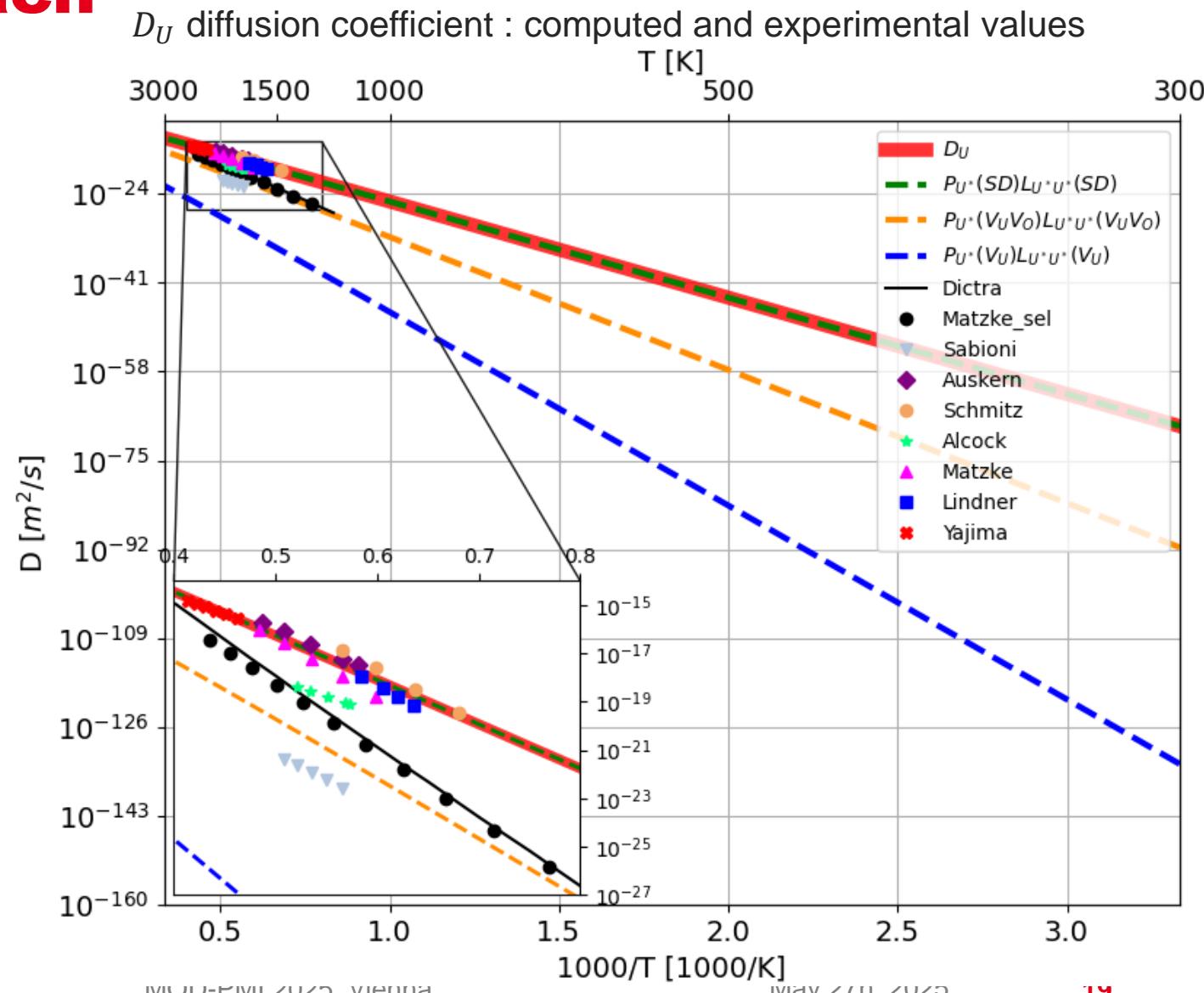
Schuler, T. Messina, L. Nastar, M. Comput. Mater. Sci. 172 (2020)

Cationic self-diffusion coefficients with combined ART + KineCluE approach

- Transport by Schottky defect is dominant
 - To be confirmed with refinement of defect model
- Satisfactory match with experiments at high temperatures
 - Thanks to several Schottky migration mechanisms found by ART and characterized by lower migration energies than isolated vacancies

D_U Arrhenian parameters	
E^A [eV]	D^0 [m^2/s]
3.65	8.41×10^{-8}

P. Ospital et al., manuscript under preparation

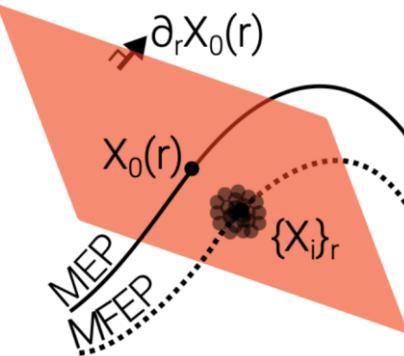




Effect of finite temperature on migration barriers with the PAFI method

T. Swinburne, M. C. Marinica, Phys. Rev. Lett. 120, 135503 (2018)

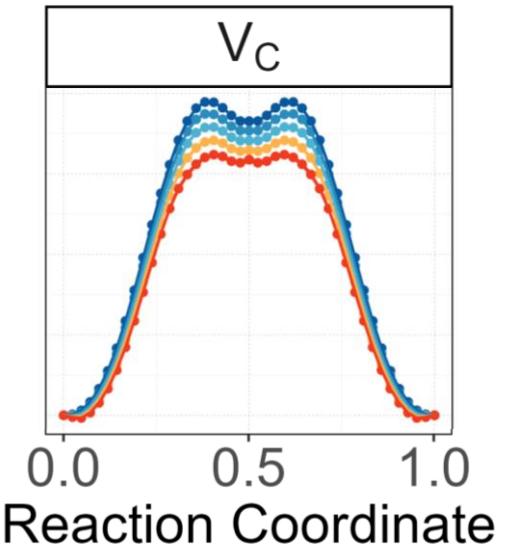
Projected Average Force Integrator



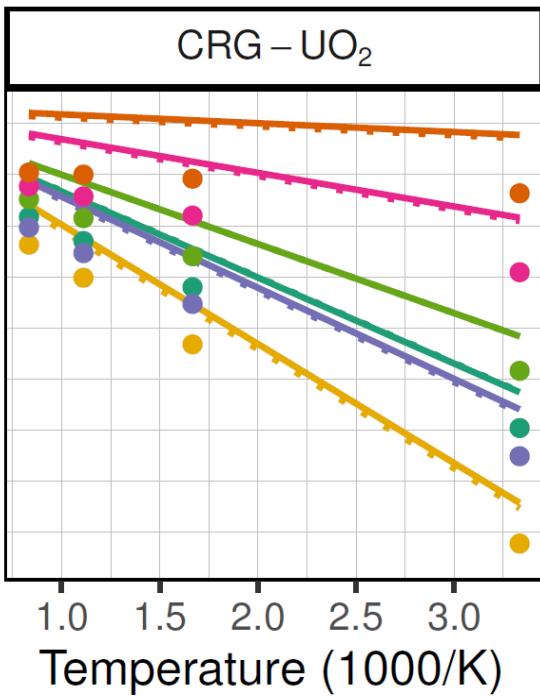
- Method to analytically determine minimum free energy path (MFEP) at finite temperature with full accounting of non-harmonic effects
- Based on adaptive biasing force method and stratified sampling of configuration space
- Starts from ground state MFEP
- Gives exact expression for free energy gradient

Cationic vacancy free energy barrier

Temperature (K) 0 300
 600 900 1200



Jump frequency vs harmonic approximation

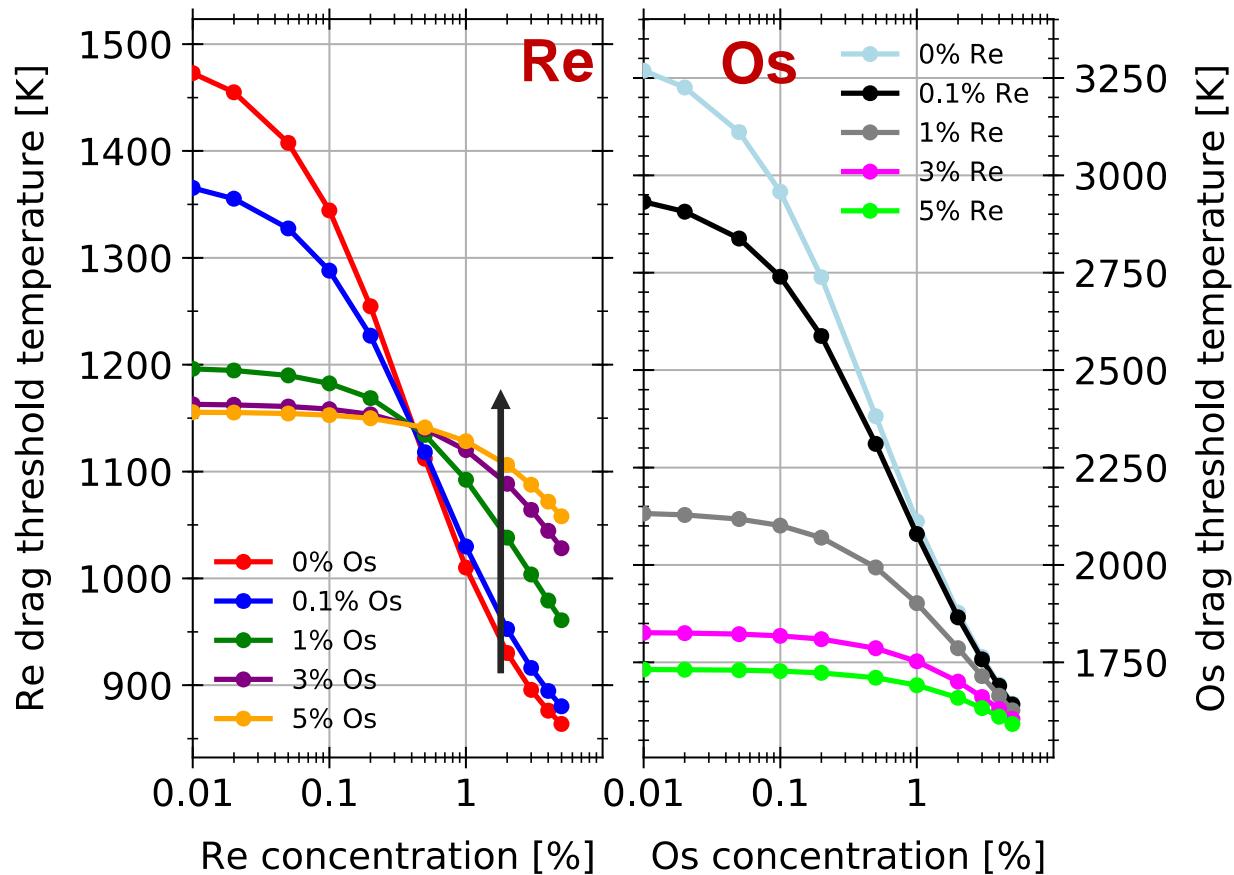


BSD	I_O
V_O	I_C
V_C	I_C^{nc}

D. Frost et al., manuscript under preparation

Kinetic coupling of defects and transmutation products in W

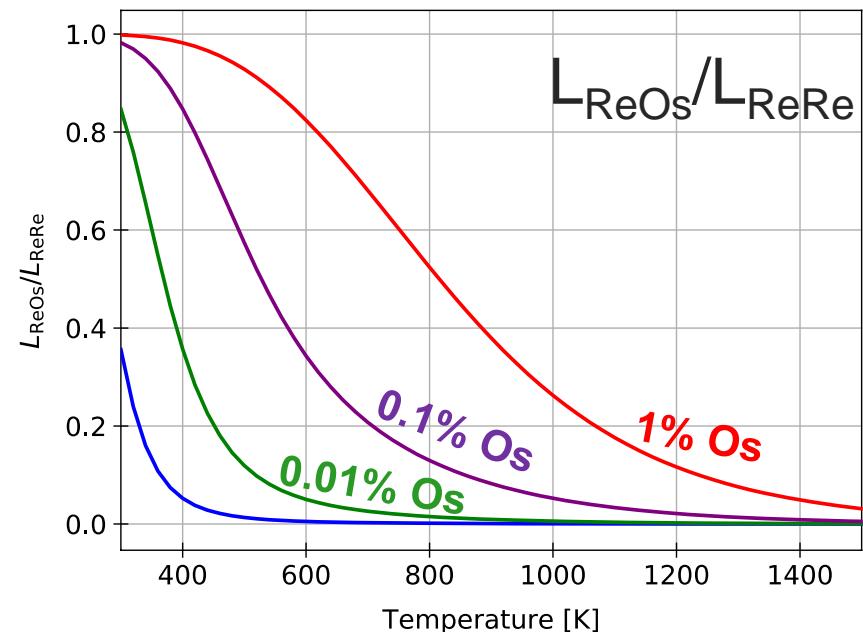
Variation of drag threshold temperature with Re/Os concentration



L. Messina, final report of EUROfusion Researcher Grant MEATBALL (2017-2019)

- Re (Os) drag weakens with increasing Re (Os) concentration
- Appearance of Os in W-Re (>0.5%Re) enhances Re drag
- Due to Re-Os coupling: as Os appears, Re diffusion can occur only through the $v_1Re_1Os_1$ cluster

Re drag by Os atoms



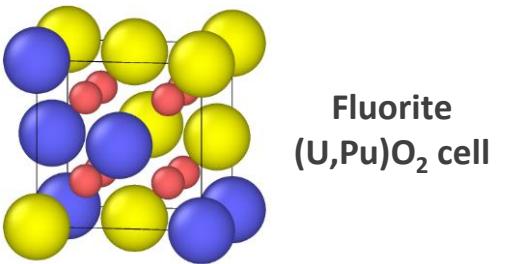


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Chemical disorder

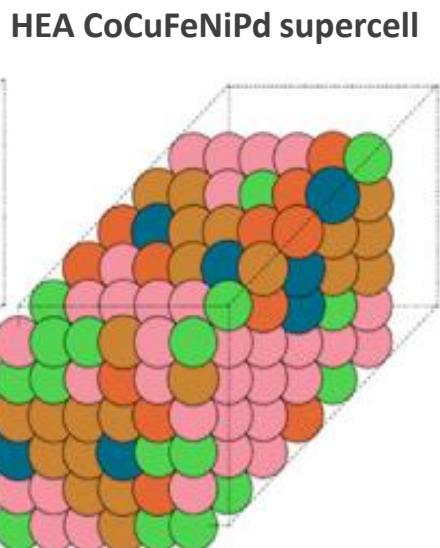
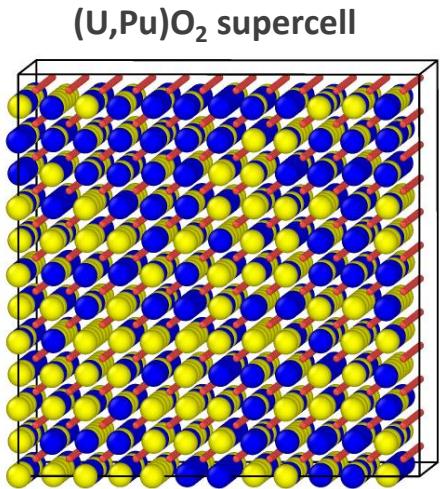
- Chemically disordered compounds: negligible mixing enthalpy
- From crystallographic point of view: multiple species sharing the same lattice/sublattice randomly (or pseudo-randomly)
- Various degrees of disorder, possible local order...
- Examples: $(U,Pu)O_2$ fuels, high-entropy alloys (HEA)



- Objective: **atomic-scale properties of disordered materials** that depend on atomic composition (i.e., distribution of chemical species on shared lattice)
- Needs **sampling of** (one or several) **partition function(s)**
- Example : point-defect equilibrium concentration:

$$C_d = \left\langle \exp\left(-\frac{G_{k,d}^{\text{form}}}{k_B T}\right) \right\rangle = \sum_{k=1}^{N_{\text{conf}}} w_k \exp\left(-\frac{G_{k,d}^{\text{form}}}{k_B T}\right)$$

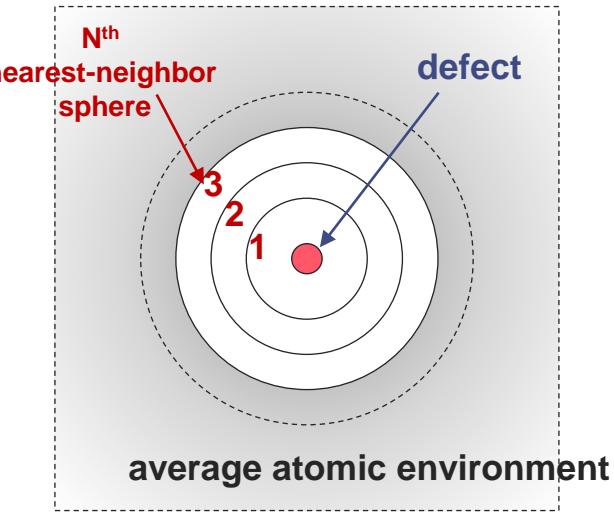
$$w_k = \frac{\exp\left(-\frac{G_k}{k_B T}\right)}{\sum_{k=1}^{N_{\text{conf}}} \exp\left(-\frac{G_k}{k_B T}\right)}$$



Limitation of traditional approaches to chemical disorder

- Brute-force approach not envisageable due to extremely large number of configurations (10^{16} with 4nn)
- Traditional approaches:
 - Special Quasi-Random Structures (SQS): generates supercells with maximized disorder, but no information on exploration exhaustivity (and not suited for locally ordered compounds)
 - Monte Carlo sampling (MCMC): demands a very high number of configurations, depending on target property
- For $(U,Pu)O_2$: very high computational cost (DFT+ U , inaccurate interatomic potentials for certain properties)
- We propose 3 approaches:
 - Systematic approach limited to nearest-neighbor atoms D. Bathellier et al., J. Appl. Phys. 132, 175103 (2022)
 - Statistical approach based on limited number of random configurations M. C. Notarangelo et al., under preparation
 - Generative AI approach to optimize exploration M. Karcz et al., Phys. Chem. Chem. Phys., 25, 23069-23080 (2023)
M. Karcz et al., arXiv:2408.14928

Generation of local atomic configurations around a central point, up to the Nth neighbor sphere

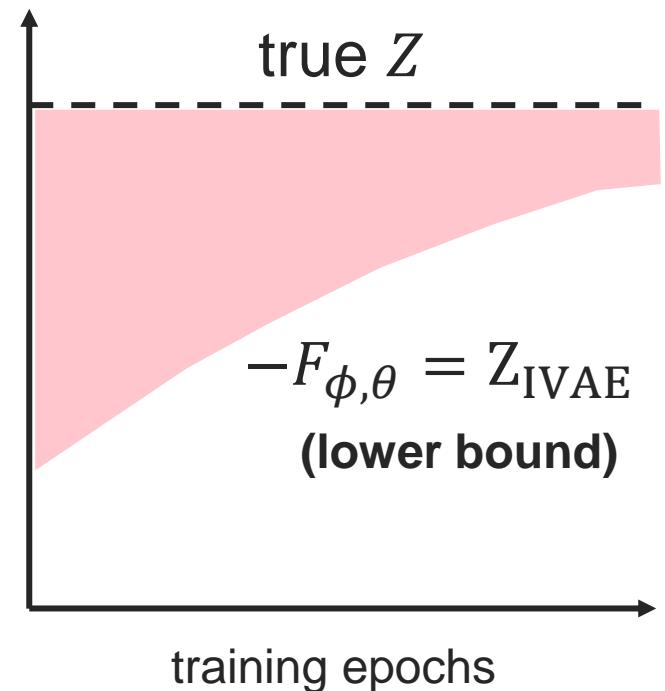
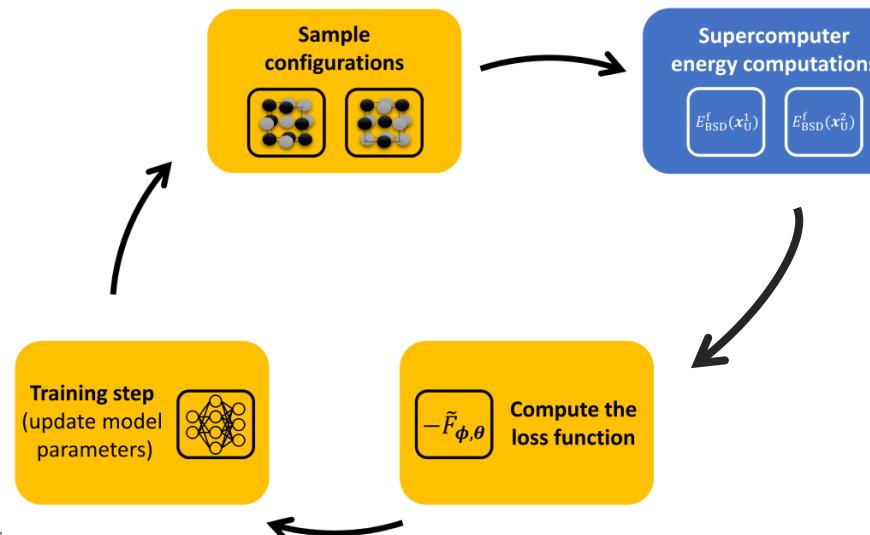


AI-driven partition function estimation and sampling: the PULSE method

PULSE : Partition function Unsupervised Learning Sampling and Evaluation for disordered compounds

- ❑ Non-supervised generative model that **builds its own training database** with an iterative **active learning approach**, with two simultaneous tasks:
 - ❑ Generation of representative atomic configurations
 - ❑ Estimation of a partition function, giving access to target property
- ❑ Requiring the lowest possible amount of training configurations (to reduce computational time needed for energy/force calculations)

M. Karcz *et al.*, arXiv:2408.14928





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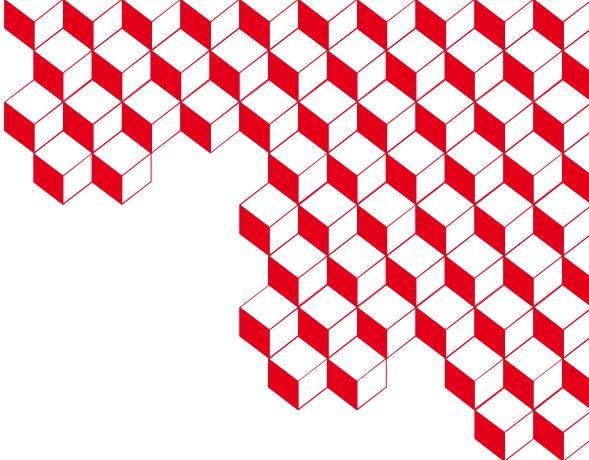
Conclusions

- Phase-field modeling of gas (He, Kr) evolution in polycrystalline microstructures
- Coupled solution of phase-field and thermomechanical equations allows for simulation of complex microstructure phenomena such as fuel restructuring
- Calculation of point defect cluster contribution to transport/diffusion coefficients by combination of ART & KineCluE
- Full non-harmonic calculation of free-energy barriers with PAFI method
- AI-based PULSE method to sample and evaluation partition function of disordered compounds

Diffusion phenomena

Inert gas behavior

High-entropy alloys & oxides



Thank you

CEA CADARACHE

13115 Saint Paul Lez Durance Cedex
France

luca.messina@cea.fr
+ 33 4 42 25 61 80

iresne@cea.fr

[Site internet CEA IRESNE](#)

[LinkedIn CEA IRESNE](#)