

Multiscale simulation and AI-driven approach for advanced materials and semiconductor processing

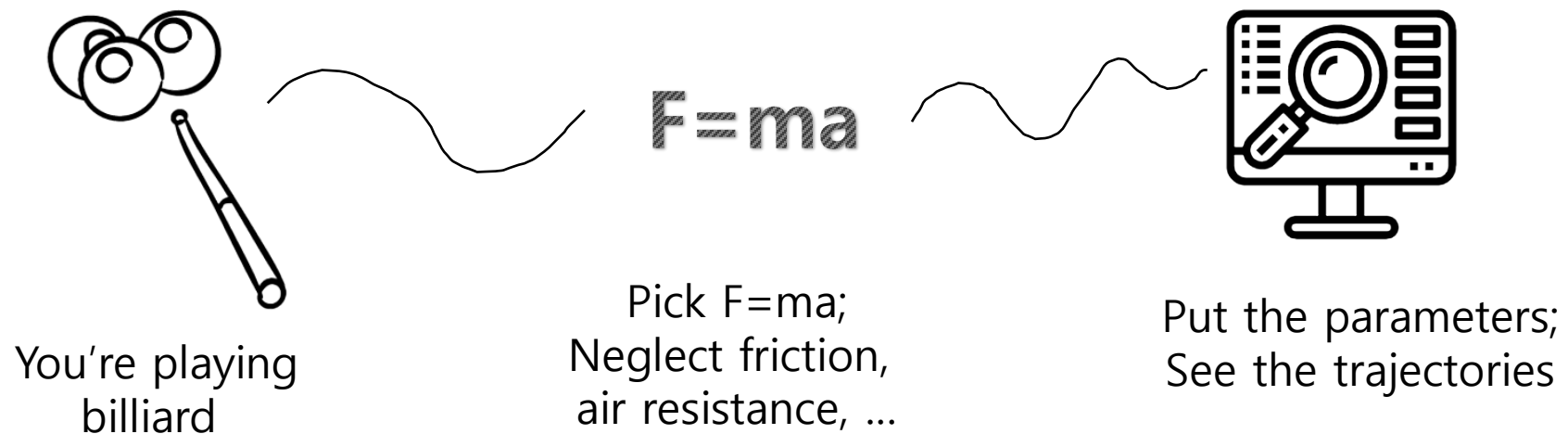
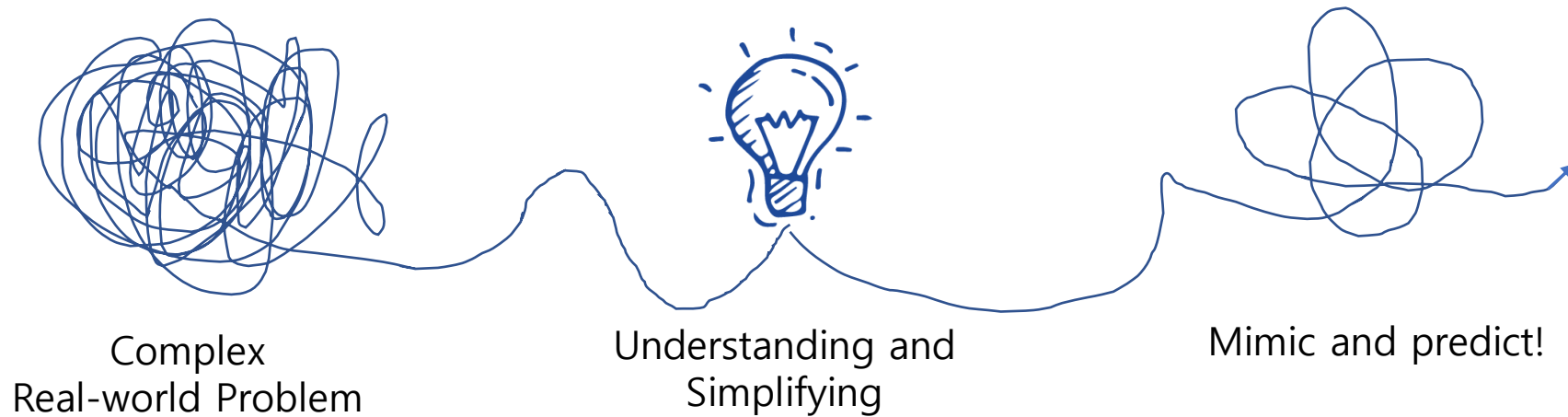
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What is modeling?



How about semiconductor process modeling?

In billiards, the trajectory from $F=ma$ is enough. However, **materials are more complex!**



(atomic)structure



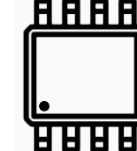
process



(micro)structure



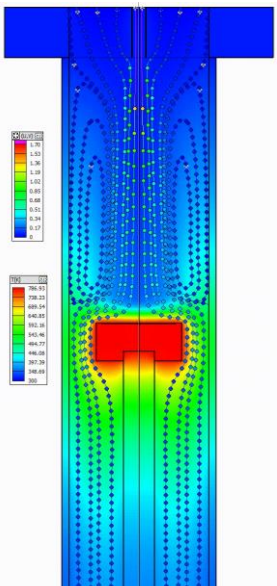
property



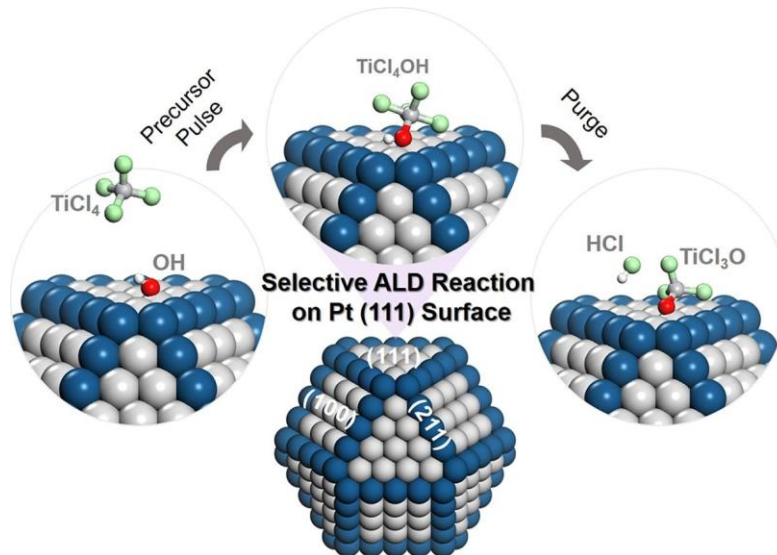
applications

It cannot be described with single equation. It's Multiscale and Multiphysics problem!

Equipment

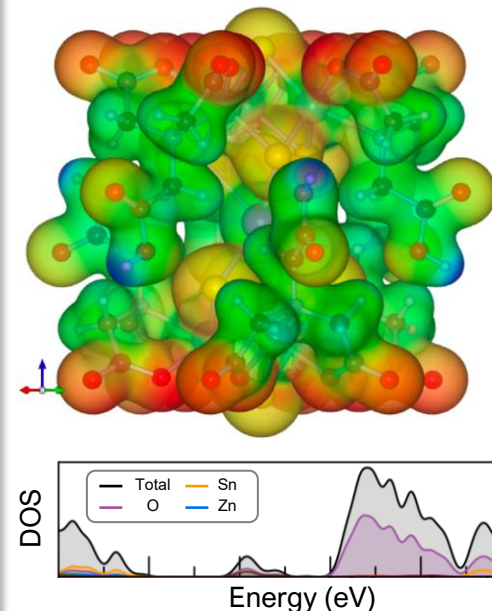


Reaction & Atomic structure

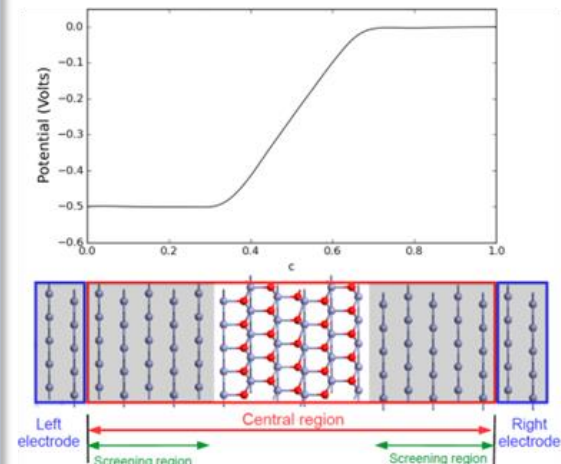


➤ 10.1016/j.apsusc.2022.154695

Electronic Structure



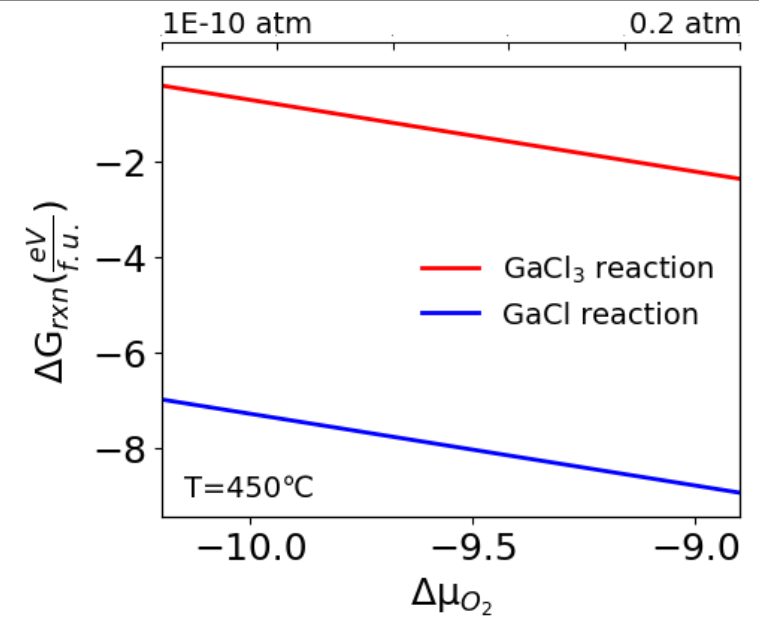
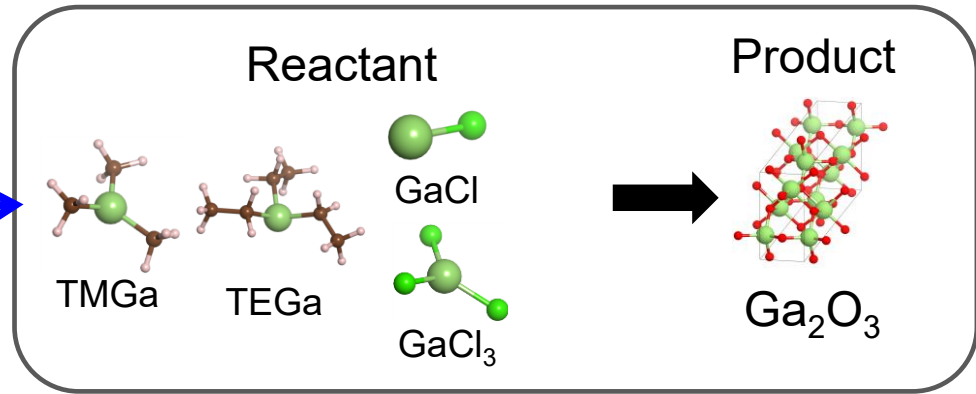
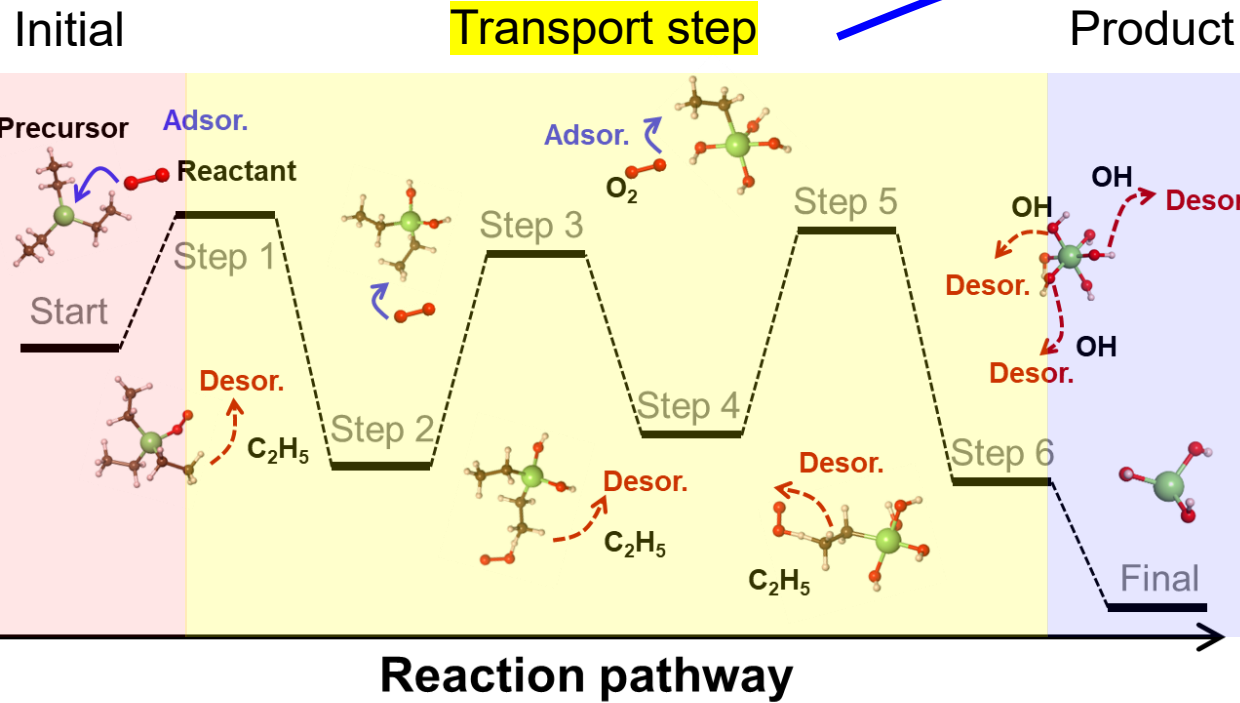
Device Property



Precursor simulations (DFT)

Decomposition process of precursors

Simplify!



$$\begin{aligned} H_1 \Psi_1 &= E_1 \Psi_1 \\ H_2 \Psi_2 &= E_2 \Psi_2 \end{aligned} \longrightarrow \Delta H_{\text{step1}}$$

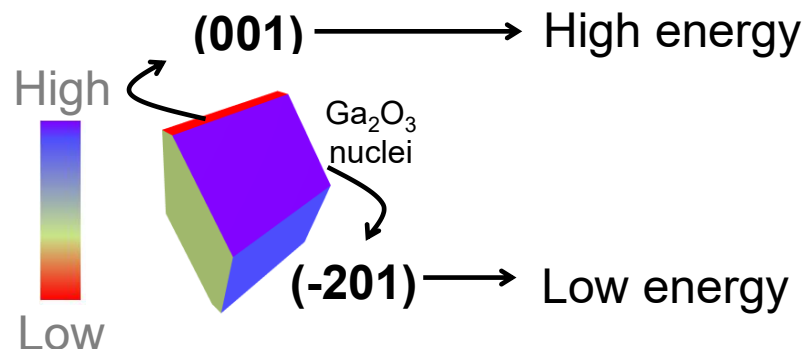
$$\begin{aligned} H_2 \Psi_2 &= E_2 \Psi_2 \\ H_3 \Psi_3 &= E_3 \Psi_3 \end{aligned} \longrightarrow \Delta H_{\text{step2}}$$

From the electronic structure,
We can extract reaction energetics w.r.t. T , p
Other information (electron polarity, etc.) is simplified

Nucleation theory simulation (DFT)

✓ Surface-dependent reaction energetics

$$G_{nuclei} = -n\Delta\mu + \gamma_e A_e$$

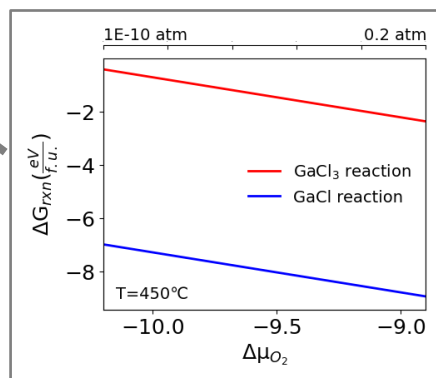


✓ Classical nucleation theory

$$Q \approx \exp\left(\frac{16\pi\gamma^3}{3k_B T (\Delta G_{rxn})^2}\right)$$

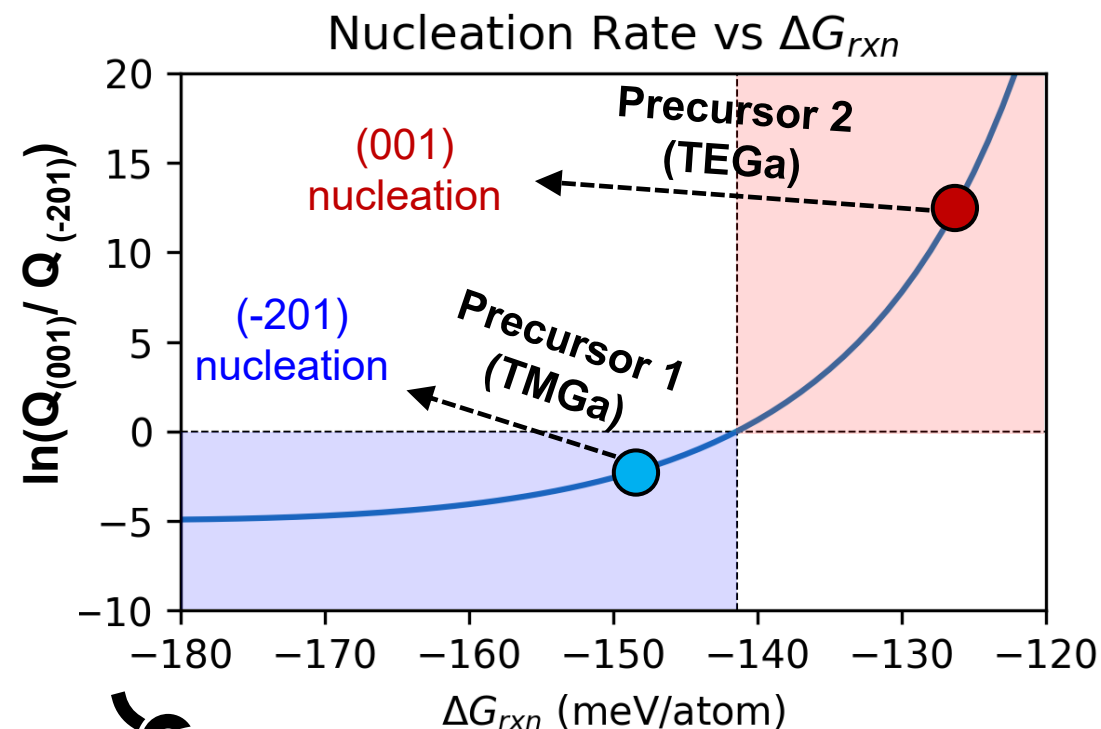
Knowledge

Reaction energy
(from precursor simulation)



nucleation, orientation, polymorph information can be extracted

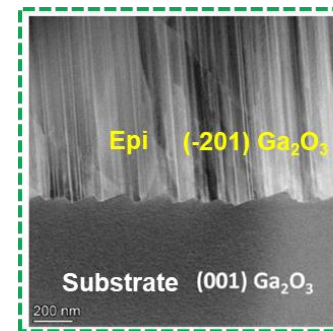
P_{O₂} = 0.21 atm, Temperature ≈ 950 °C



Exp.

Using TMGa

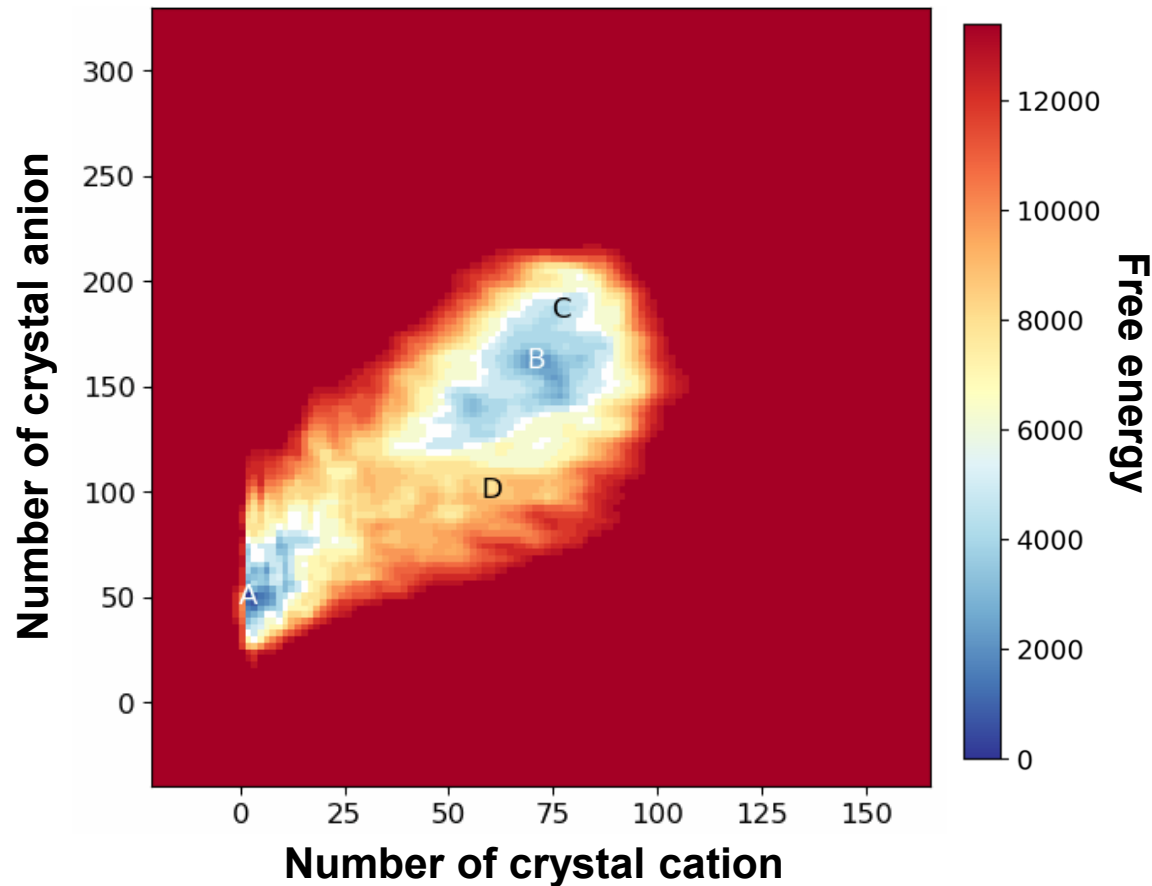
(-201) growth!



Nucleation kinetics analysis (MD)

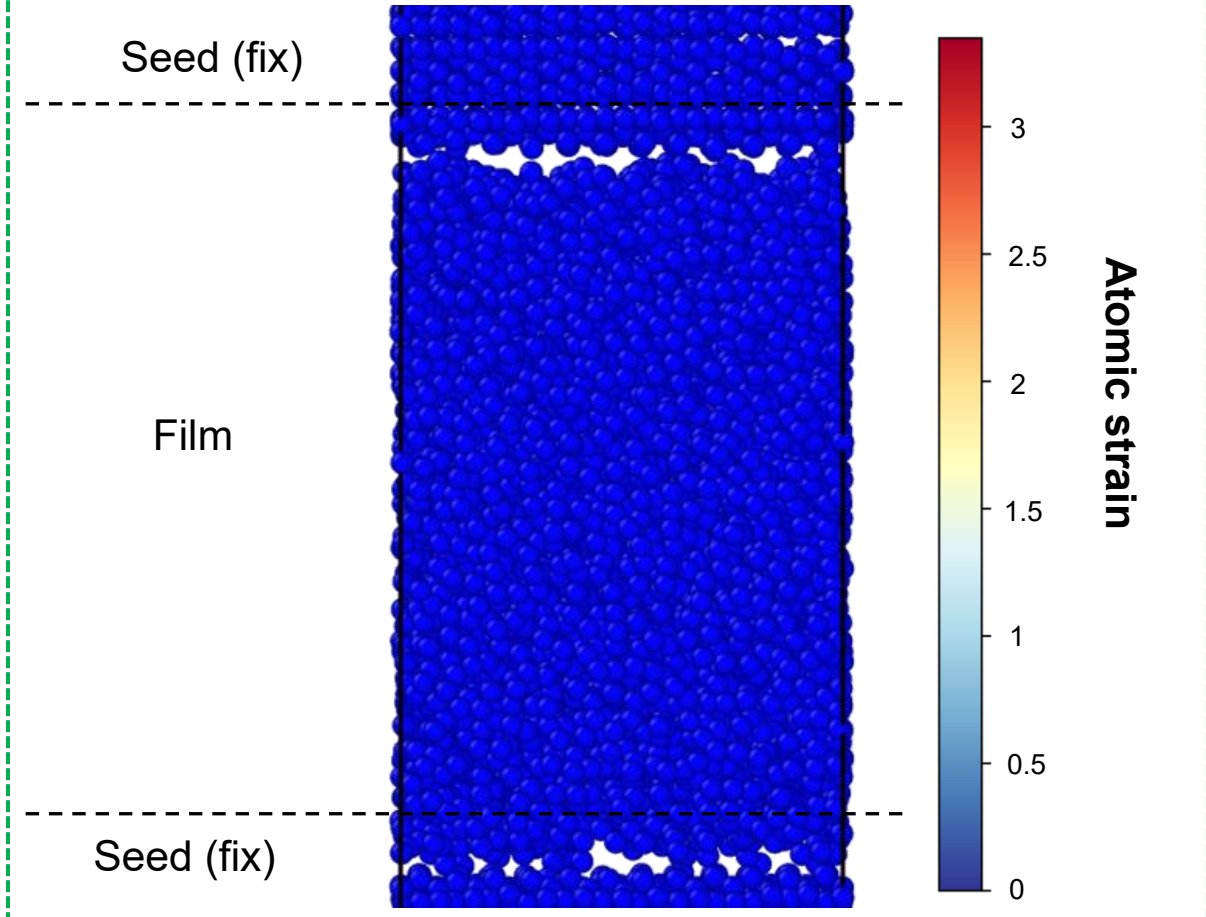
1. Nucleation free energy

Exploring nucleation path (0 ns $\rightarrow$ 10 ns)

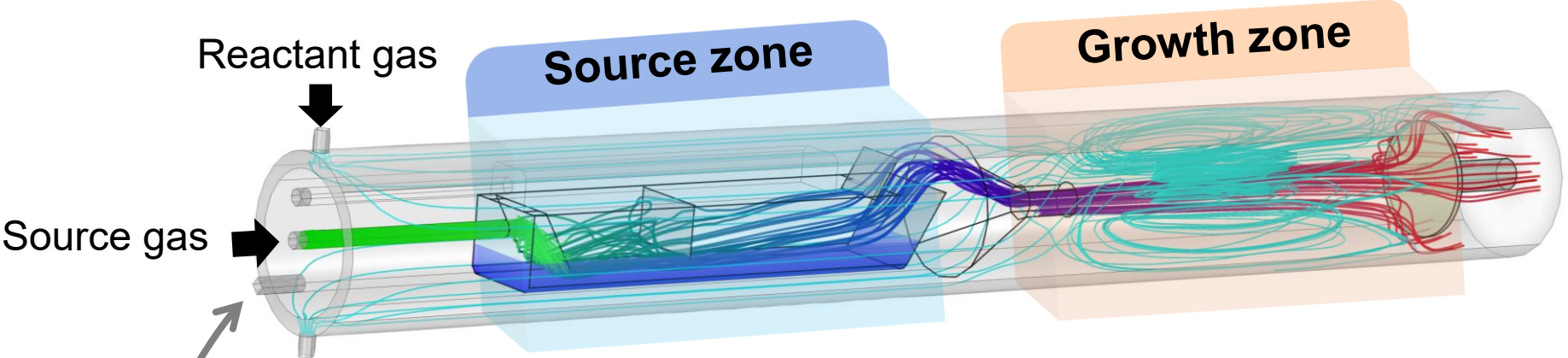


2. Interfacial strain analysis

Growth simulation (0 ns $\rightarrow$ 10 ns)



HVPE chamber modeling based on Multiphysics FEM

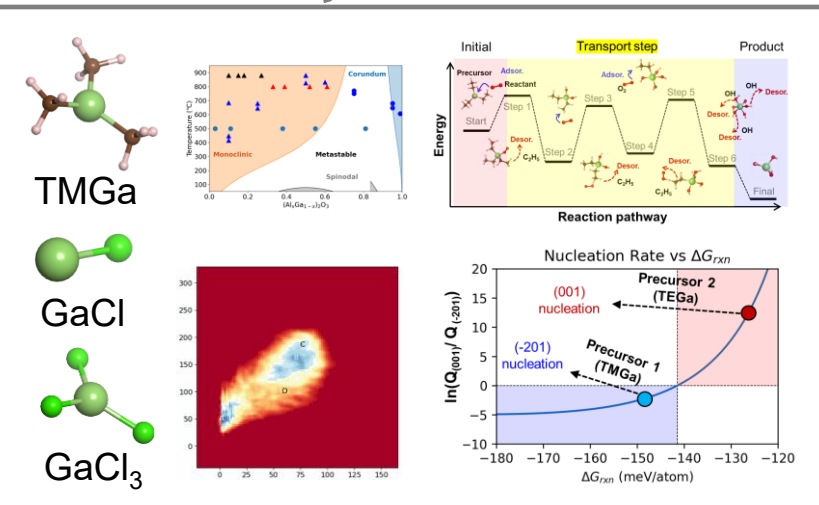


Governin Eq.

Fluid dynamics
Heat transport
Gas diffusion

Parameterized Knowledge

Variable call-up



✗ Non-selectable

✓ Selectable

- ✓ Gas flow (N₂, sccm)
- ✓ Susceptor position (cm)
- ✓ Temperature (°C)

- ✓ Turbulence and vortex
- ✓ Molecule diffusion
- ✓ Gas pressure
- ✓ ...

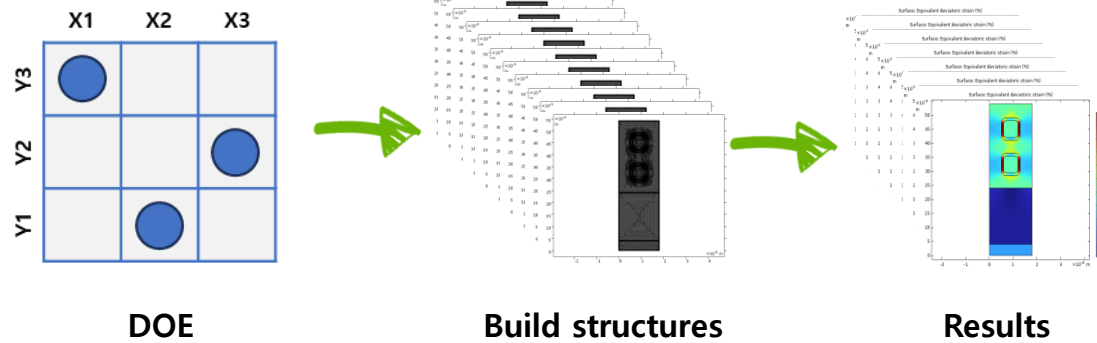
Fluid-related

- ✓ Temperature gradient
- ✓ Gas temperature
- ✓ Nucleation Reactions
- ✓ ...

Thermal-related

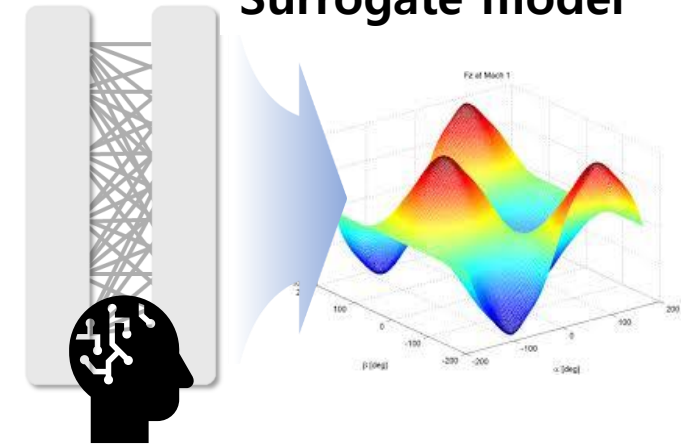
Optimizing process

Parametric study (data collection)



AI modeling

Surrogate model



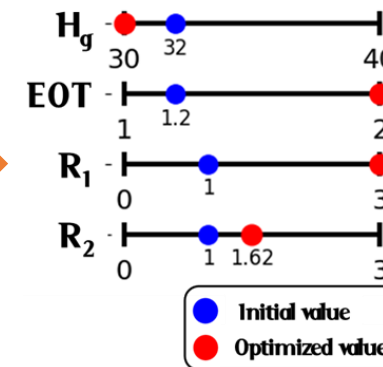
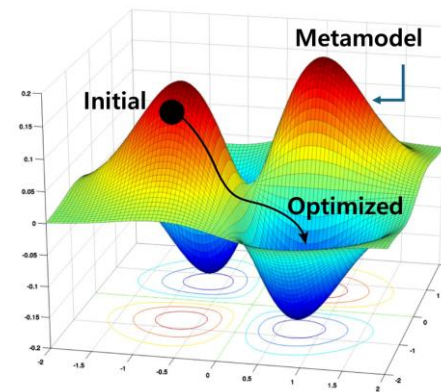
Optimization

GEOMETRY PARAMETERS

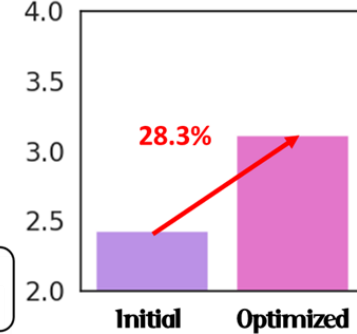
- R_1 : 1-3
- R_2 : 1-3
- EOT : 1-2
- H_g : 30-40

TARGET PROPERTY

- Effective mass : Minimizing



Effective Mass Reduction(%)

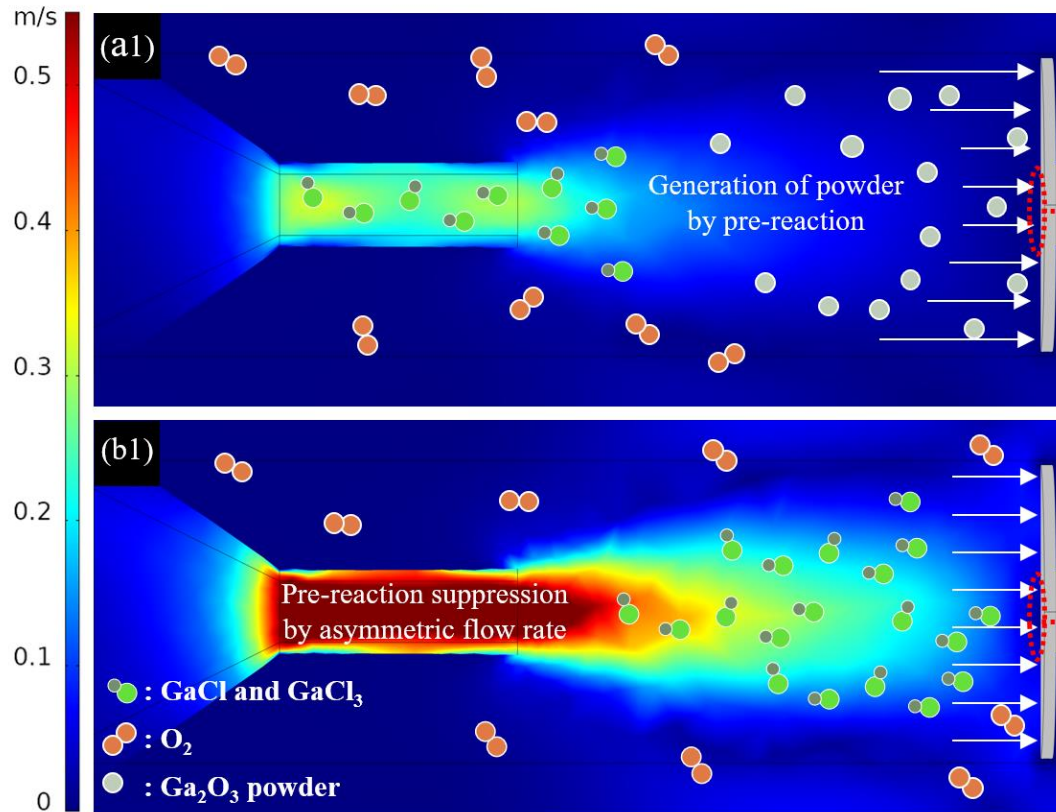


Experimental validation on deposition quality

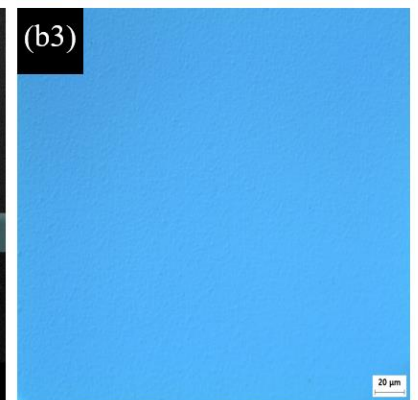
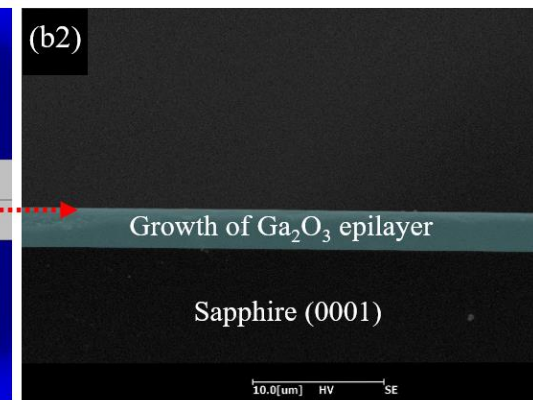
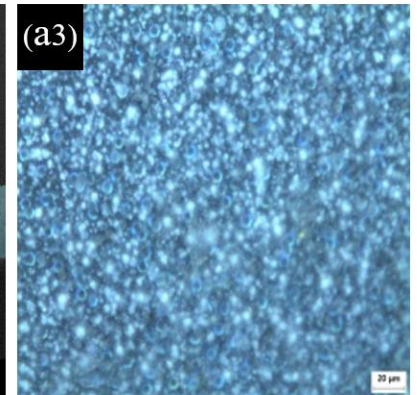
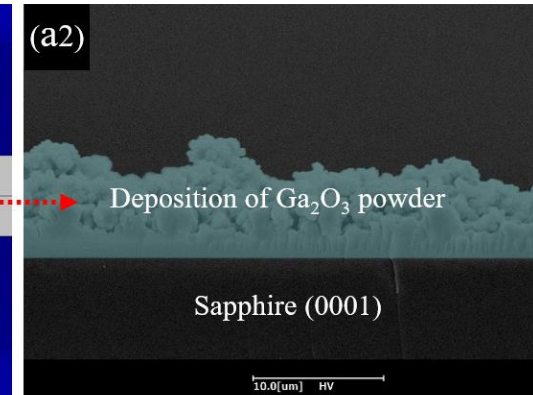
Computation



Optimization!



Gas velocity in growth zone



SEM

OM image

Concluding Remarks

- The key idea of modeling is capturing essential physics in complex by simplifying the phenomena
- For real problem, we need to solve multiple governing equations in multiscale & multiphysics dimension
- We have constructed AI-assisted simulator to get optimized recipe

