

# Current and Future Perspectives on Atomic Layer Deposition

Erwin Kessels



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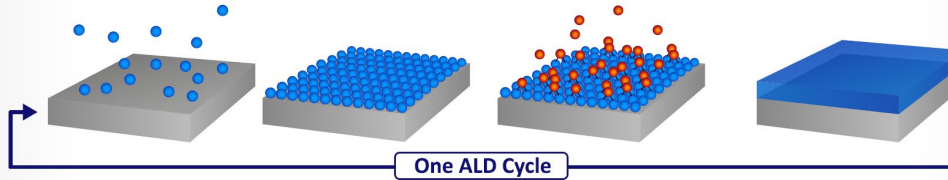
erwin-kessels

# Atomic scale processing toolbox

Providing complementary routes to build, remove, and pattern materials atom by atom

## ALD

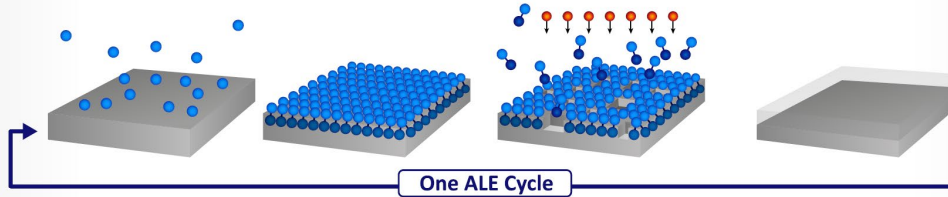
Atomic Layer Deposition



- Thermal ALD
- Plasma-enhanced ALD

## ALE

Atomic Layer Etching

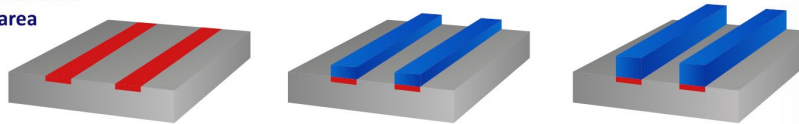


|             | PLASMA | THERMAL |
|-------------|--------|---------|
| Isotropic   | ✓      | ✓       |
| Anisotropic | ✓      | ✗       |

## ASD

Area-Selective Deposition

■ non-growth area  
■ growth area



- Intrinsic
- Activation
- Deactivation
  - Self-assembled monolayers
  - Small molecule inhibitors

# Outline

1. **Quick introduction to ALD**
2. ALD process development
3. Towards AI-enabled ALD
4. Challenges in frontier materials

# Recent review paper on ALD – Nature Reviews Methods Primers

nature reviews methods primers

<https://doi.org/10.1038/s43586-025-00435-6>

Primer

 Check for updates

## Atomic layer deposition

Erwin Kessels<sup>1</sup>✉, Anjana Devi<sup>2,3,4</sup>, Jin-Seong Park<sup>5</sup>, Mikko Ritala<sup>6</sup>, Angel Yanguas-Gil<sup>7</sup> & Claudia Wiemer<sup>8</sup>

Abstract

Atomic layer deposition (ALD) is a surface-controlled chemical vapour deposition method, in which materials are prepared one atomic layer at the time. With ALD, film thickness can be controlled very precisely, and it allows the user to cover large areas and surfaces with a complex three-dimensional structure uniformly and conformally. ALD is used for the deposition of high-quality thin films and nanostructures, as well as for surface functionalization and interface engineering in a wide range of applications, both from a research and development perspective, as well as for high-volume manufacturing. This Primer

Sections

Introduction

Experimentation

Results

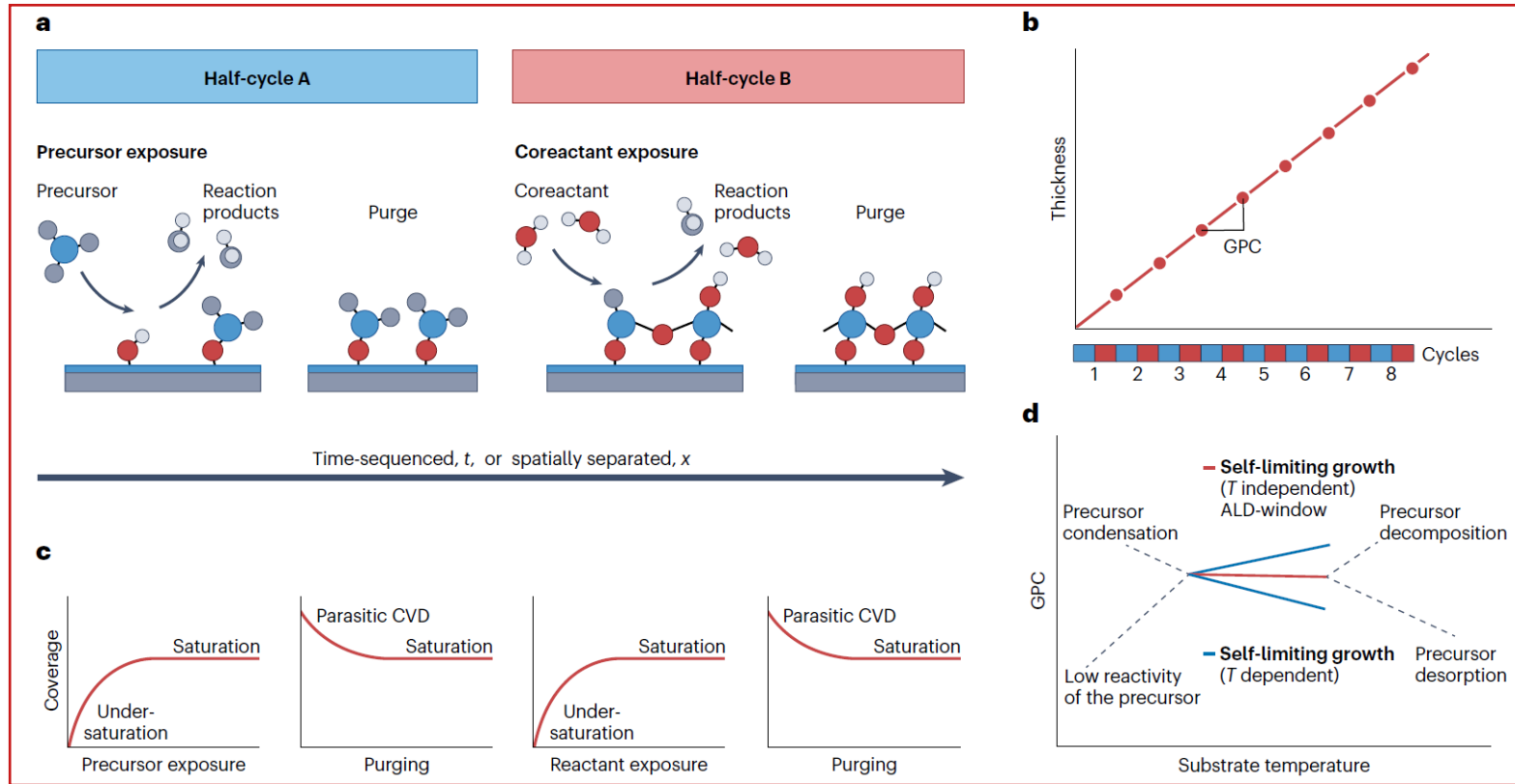
Applications

Reproducibility and  
data deposition

Limitations and optimizations

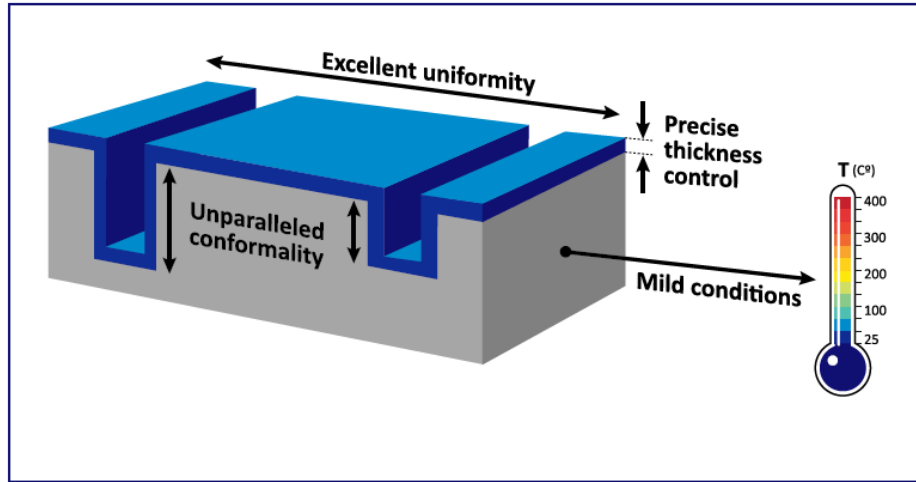
Outlook

# ALD in one slide: self-limiting cycles enable precise growth

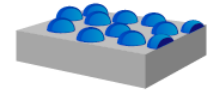


# ALD: enabling nanotechnology through precision and conformality

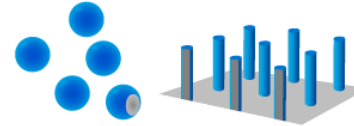
ALD enables controlled materials synthesis on complex surfaces and nanostructures



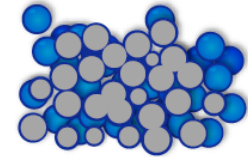
*Flexible foils*



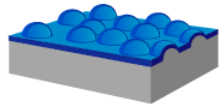
*Size-controlled nanoparticles*



*Conformal coating of nanoparticles and nanowires*



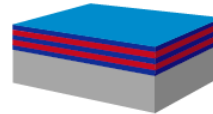
*Conformal penetration of porous substrates*



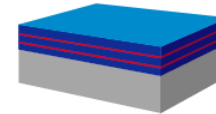
*Thin film encapsulation*



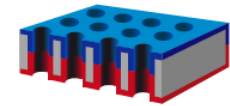
*Selective deposition on prepatterned surfaces*



*Nanolaminates*



*Delta-doping*

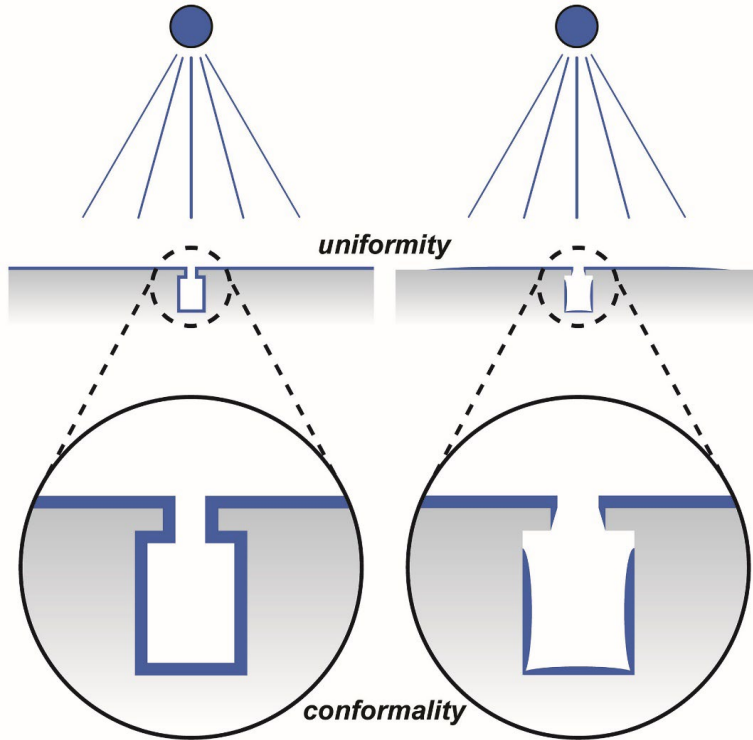


*Nanoscale engineering*

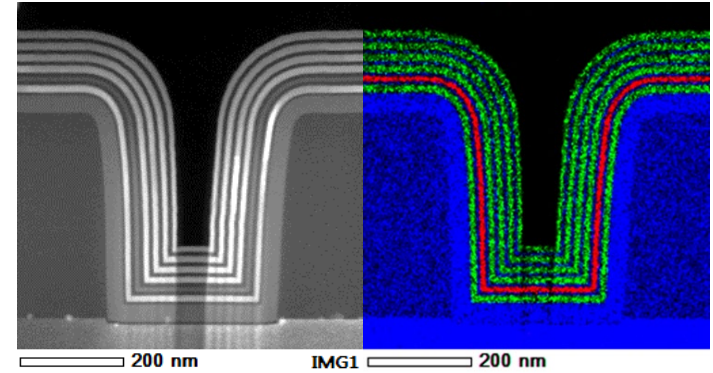
# ALD conformality comes from surface-controlled growth

(a) **Surface controlled**

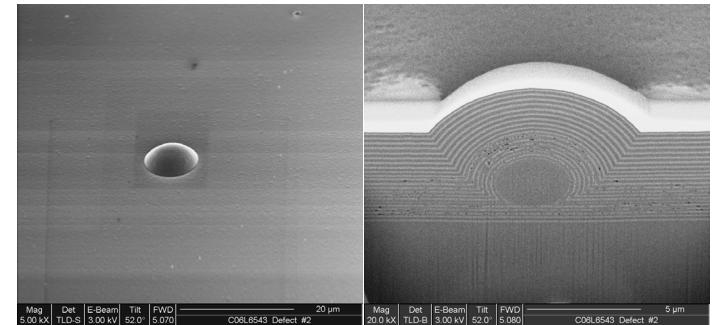
(b) **Flux controlled**



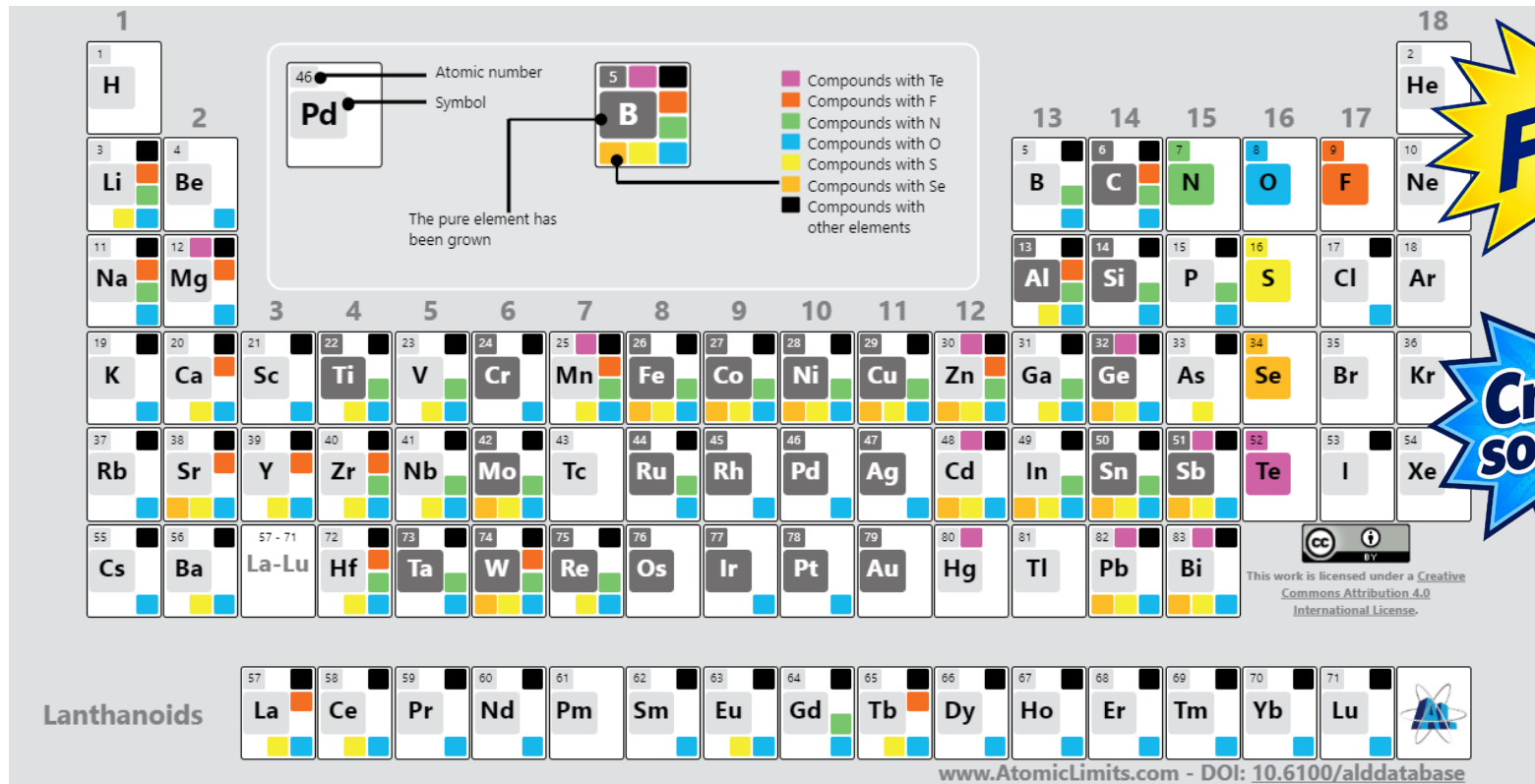
Film stack in etched trench (Arts *et al.*)



Encapsulated particle on wafer (Jim Wang)



# AtomicLimits ALD Database: all published processes with refs.



# ALD is simple in principle, but rich in choices

## ALD precursors

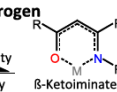
Coordination via:

### ► Oxygen

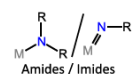


### ► Oxygen & Nitrogen

Increasing Reactivity  
Increasing Stability



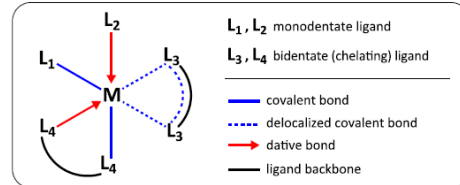
### ► Nitrogen



### ► Carbon



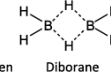
### ► Others



## ALD co-reactants

Material class:

### ► Metals



Plasmas of:  $H_2, O_2$

### ► Nitrides



Plasmas of:  $H_2, N_2, NH_3$

### ► Oxides



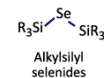
Plasmas of:  $O_2, N_2O$

### ► Sulfides

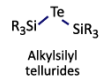


Plasmas of:  $H_2S$

### ► Selenides



### ► Tellurides



## ALD reactors

### ► Temporal ALD

- Flow-type
- Showerhead
- Batch
- Direct plasma
- Remote plasma
- ...

### ► Spatial ALD

- Sheet-to-sheet
- Roll-to-roll
- Rotating semi-batch

### ► Particle ALD

- Fluidized-bed
- Rotating drum
- ...

## ALD cycles

### ► AB-type

A B A B A B A B

### ► Advanced ABC-type

A B C A B C A B C A B C

### ► Supercycles $(AB)_m(CD)_n$

A B C D A B C D A B C D A B C D

## ALD parameters

### ► Doses and exposures (time – partial pressure)

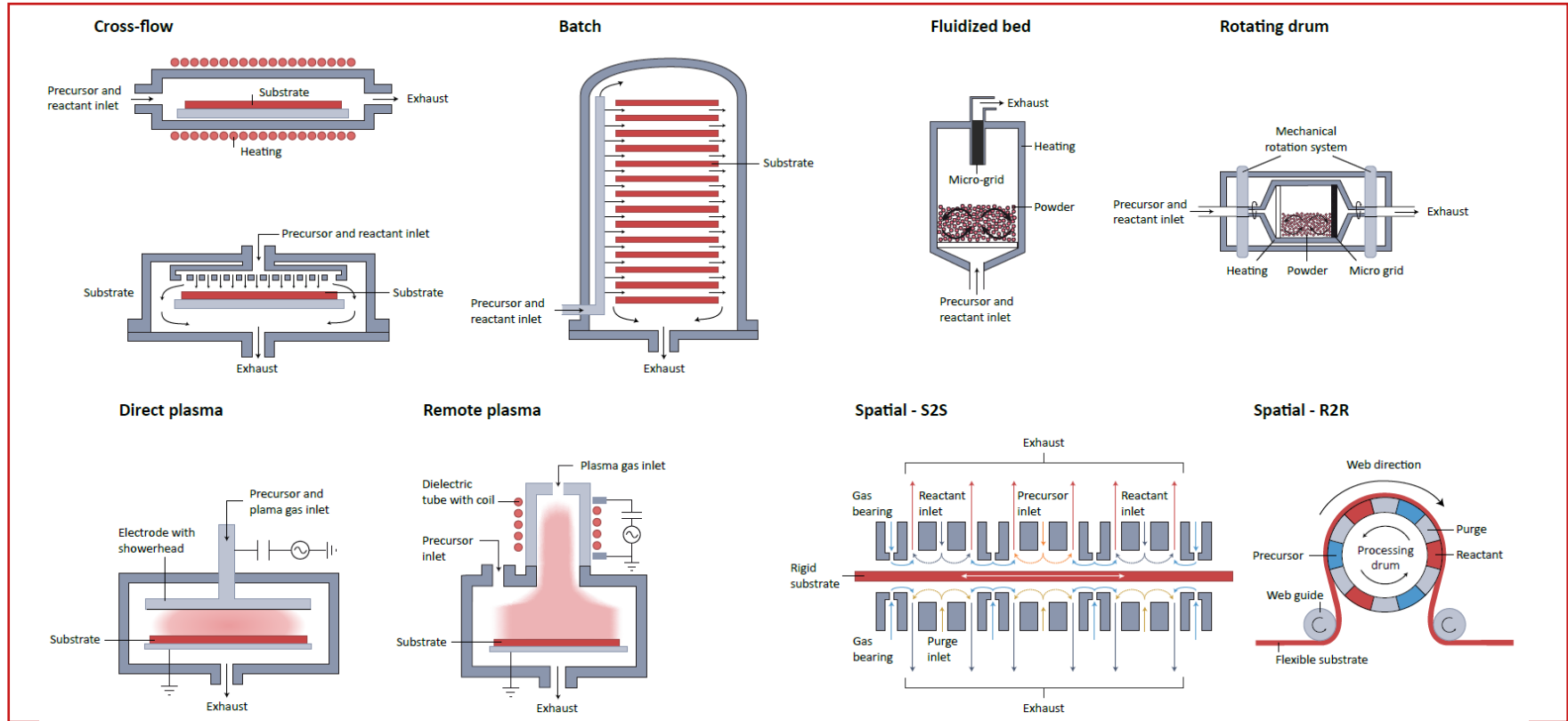
### ► Temperature(s)

### ► Pressure

### ► Purge gases

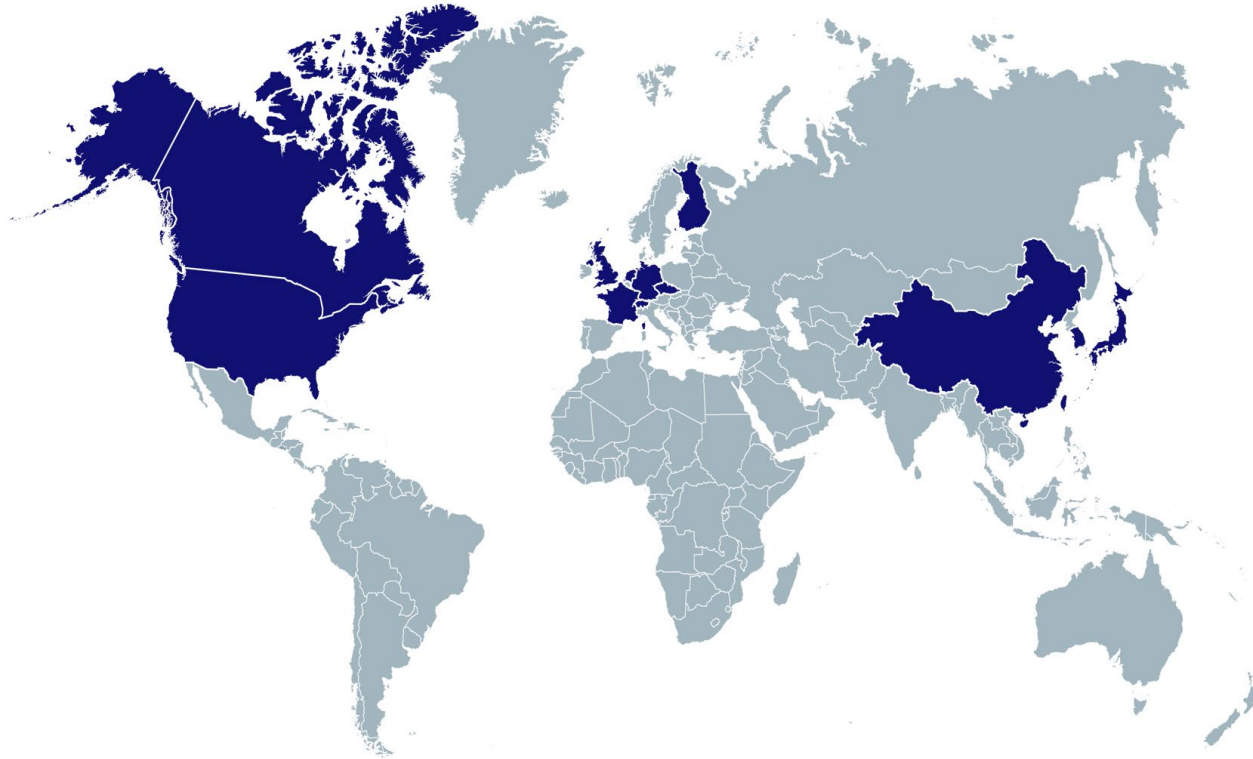
### ► Etc.

# One ALD principle, many reactor architectures



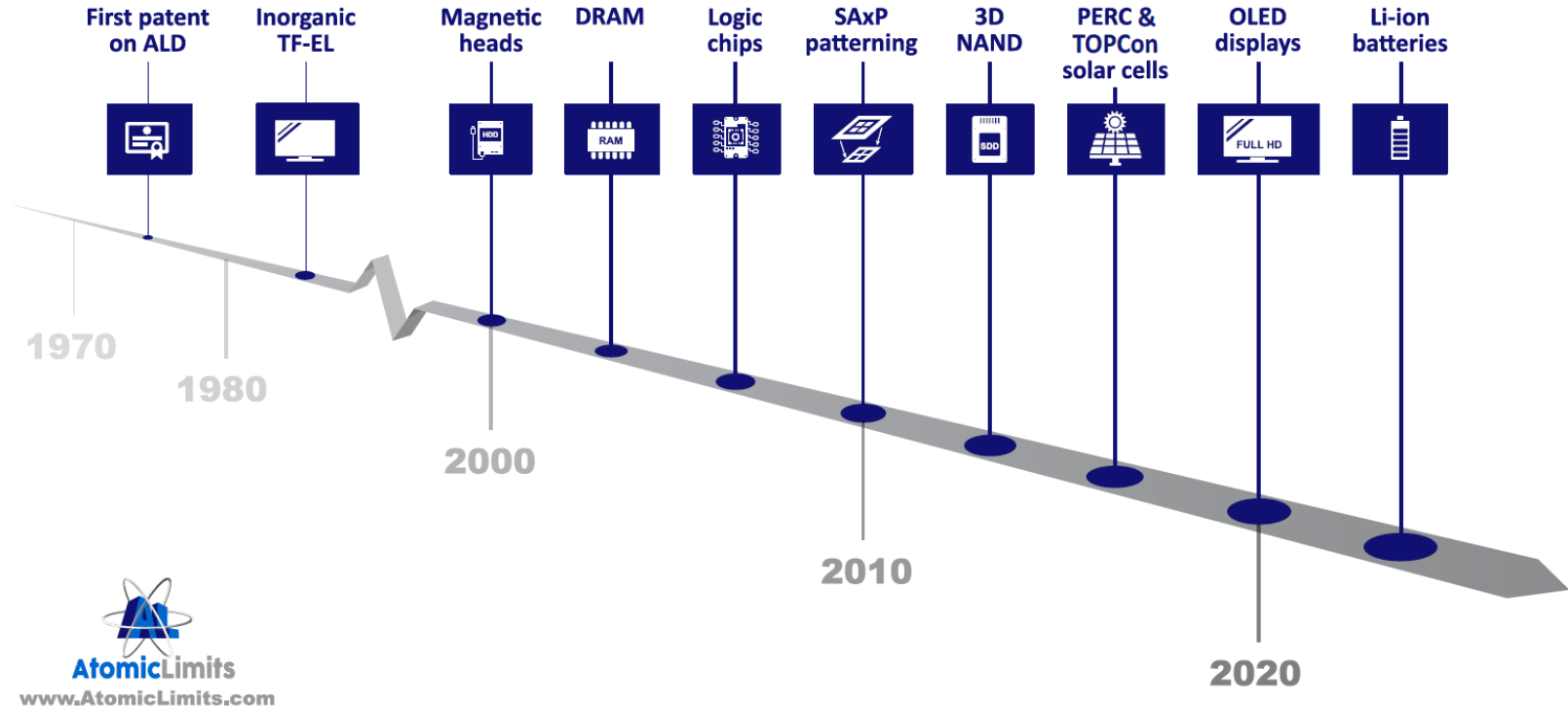
# The global equipment landscape for ALD – 60+ companies

Animation

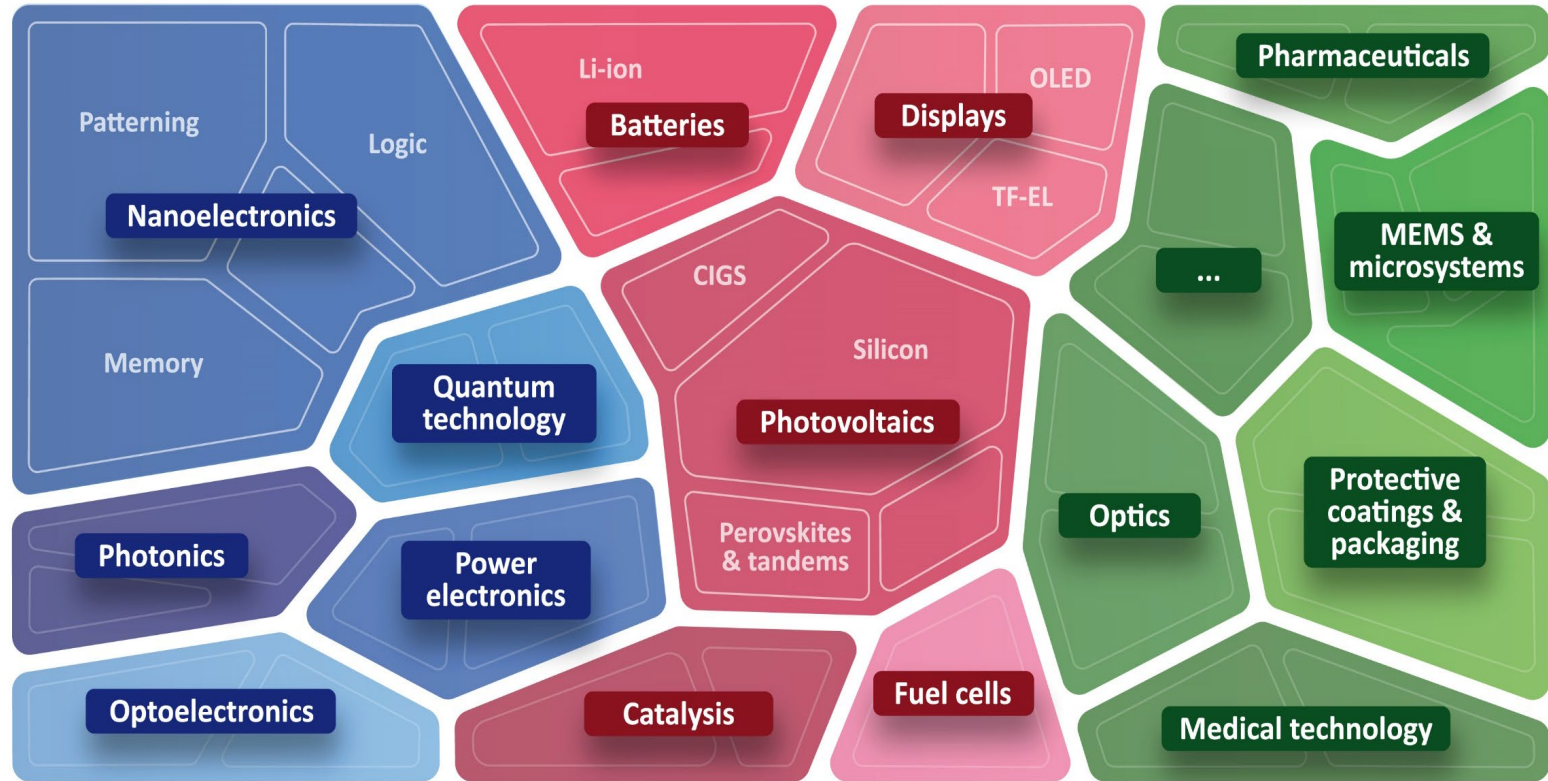


# From invention to industrial impact: 50+ years of ALD

## Applications of ALD in high-volume manufacturing (HVM)



# ALD enables a broad landscape of applications



# AtomicLimits ReviewBase: Browse all review papers

AtomicLimits ReviewBase

Search ...

ALD ASD Filter & sort

4927 citations

Atomic Layer Deposition: An Overview

Steven M. George, *Chem. Rev.* **110**, 111- 131 (2009)

2280 citations

Surface chemistry of atomic layer deposition: A case study for the trimethylaluminum/water process

Riikka L. Puurunen, *Journal of Applied Physics* **97**, (2005)

1571 citations

A brief review of atomic layer deposition: from fundamentals to applications

Richard W. Johnson, Adam Hultqvist, Stacey F. Bent, *Materials Today* **17**, 236- 246 (2014)

1257 citations

Crystallinity of inorganic films grown by atomic layer deposition: Overview and general trends

Ville Miikkulainen, Markku Leskelä, Mikko Ritala, Riikka L. Puurunen, *Journal of Applied Physics* **113**, (2013)

1126 citations

Atomic layer deposition (ALD): from precursors to thin film structures

Markku Leskelä, Mikko Ritala, *Thin Solid Films* **409**, 138- 146 (2002)

977 citations

Atomic Layer Deposition Chemistry: Recent Developments and Future Challenges

Markku Leskelä, Mikko Ritala, *Angew Chem Int Ed* **42**, 5548- 5554 (2003)

823 citations

Synthesis and Surface Engineering of Complex Nanostructures by Atomic Layer Deposition

M. Knez, K. Nielsch, L. Niinistö, *Advanced Materials* **19**, 3425- 3438 (2007)

749 citations

Plasma-Assisted Atomic Layer Deposition: Basics, Opportunities, and Challenges

H. B. Profijt, S. E. Potts, M. C. M. van de Sanden, W. M. M. Kessels, *Journal of Vacuum Science & Technology A: Vacuum, Surfaces, and Films* **29**, (2011)

696 citations

Catalyst Design with Atomic Layer Deposition

Brandon J. O'Neill, David H. K. Jackson, Jechan Lee, Christian Canlas, Peter C. Stair, Christopher L. Marshall, Jeffrey W. Elam, Thomas F. Kuech, James A. Dumesic, George W. Huber, *ACS Catal.* **5**, 1804- 1825 (2015)

1 like

1 like

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EQP for analysis of reaction kinetics of plasma assisted ALD processes

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TU/e

# Outline

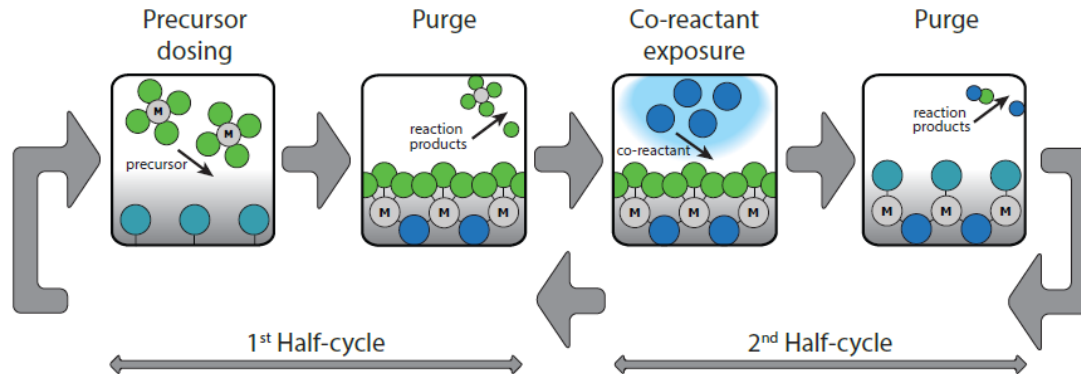
1. Quick introduction to ALD
2. **ALD process development**
3. Towards AI-enabled ALD
4. Challenges in frontier materials

# Blog post & document – AtomicLimits.com

## Atomic Layer Deposition Process Development

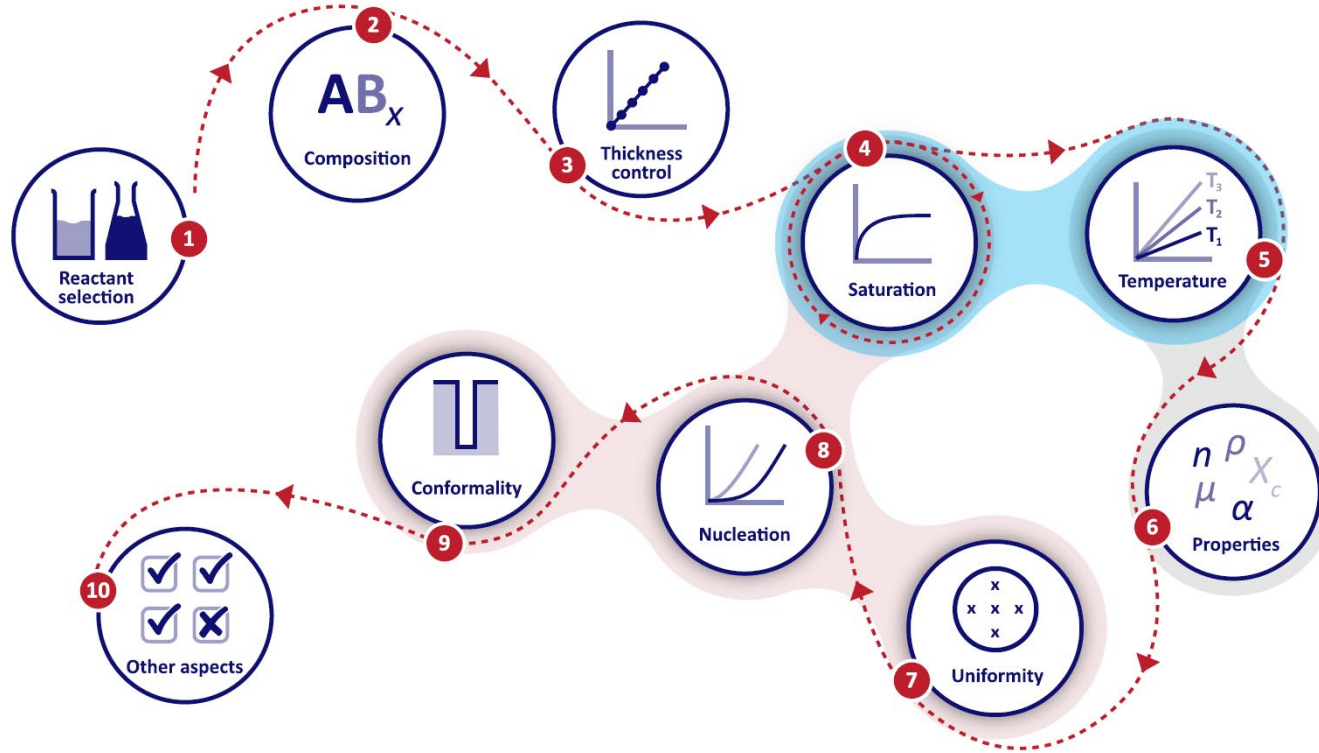
10 steps to successfully develop, optimize and characterize ALD recipes

Atomic layer deposition (ALD) is a surface-controlled thin film deposition technique that can enable ultimate control over the film thickness, uniformity on large-area substrates and conformality on 3D (nano)structures.<sup>1-5</sup> Each ALD cycle consists at least two half-cycles (but can be more complex), containing a precursor dose step and a co-reactant exposure step, separated by purge or pump steps (see Figure 1). Ideally the same amount of material is deposited in each cycle, due to the self-limiting nature of the reactions of the precursor and co-reactant with the surface groups on the substrate. By carrying out a certain number of ALD cycles, the targeted film thickness can be obtained.



# ALD process development: a structured 10-step workflow

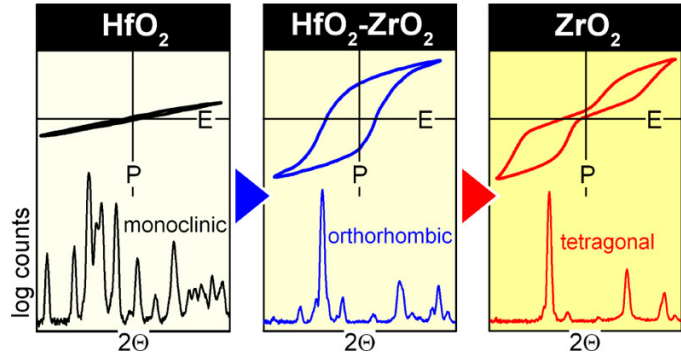
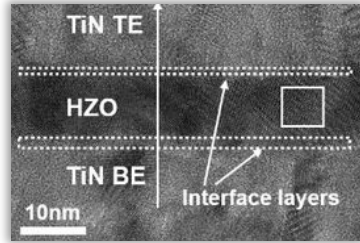
Steps ordered roughly chronologically, with iteration often needed



# ALD process development for $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ – Illustrative example

An illustrative case study to demonstrate the process-development workflow

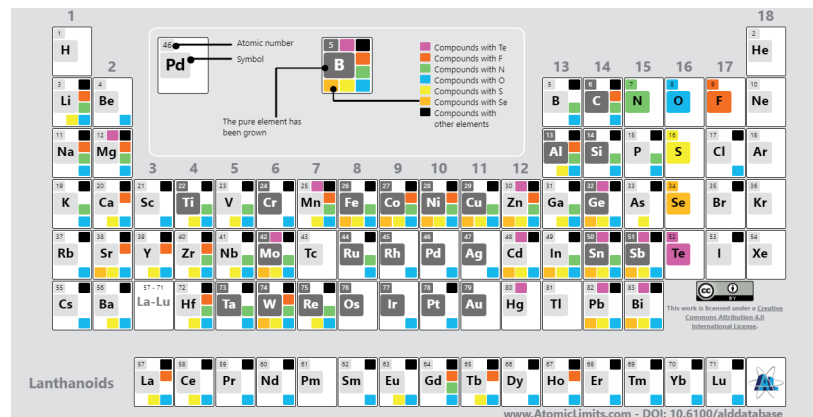
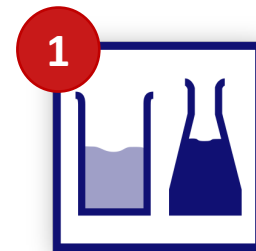
$\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$  (HZO)  
with  
ferroelectricity



| ALD development step | $\text{HfO}_2$ | $\text{ZrO}_2$ | $\text{Hf}_{1-x}\text{Zr}_x\text{O}_2$ |
|----------------------|----------------|----------------|----------------------------------------|
| 1 Reactant selection |                |                |                                        |
| 2 Composition        |                |                |                                        |
| 3 Thickness control  |                |                |                                        |
| 4 Saturation         |                |                |                                        |
| 5 Properties         |                |                |                                        |
| 6 Temperature        |                |                |                                        |
| 7 Uniformity         |                |                |                                        |
| 8 Conformality       |                |                |                                        |
| 9 Nucleation         |                |                |                                        |
| 10 Other aspects     |                |                |                                        |

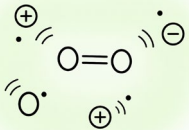
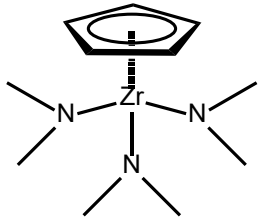
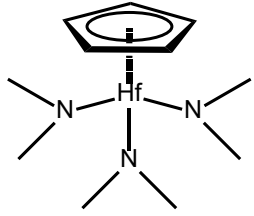
# Step 1: Reactant selection - Best choice?

- Right precursor/co-reactant pair is critical
- Must deliver the required elements to the surface
- Need to balance volatility, stability, and reactivity
- Must fit reactor and process constraints
- Literature (ALD database) provides the first shortlist

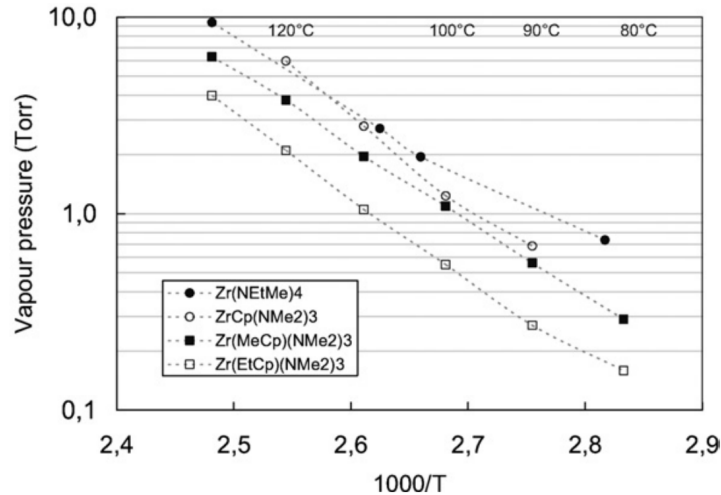


# Step 1: Reactant selection - $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

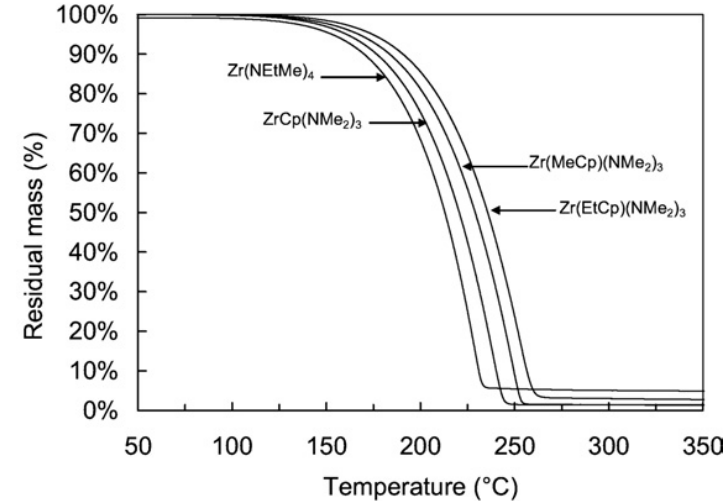
1

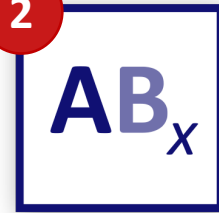


Vapor pressure for Zr precursors



Thermogravimetric analysis (TGA)





## Step 2: Composition - Desired material?

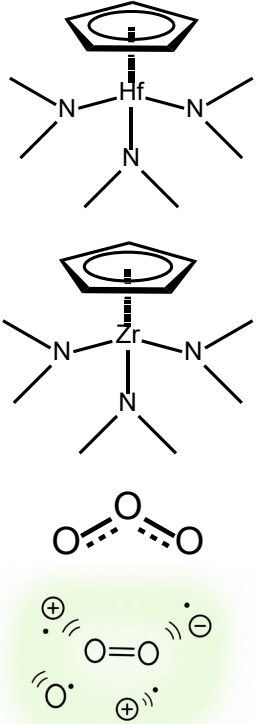
- Check the material after an early test run
- Start with fast, accessible diagnostics
  - Refractive index by ellipsometry
  - Composition by XPS
  - Conductivity for metallic films, etc.
  - ...
- If results strongly deviate, revisit reactant selection



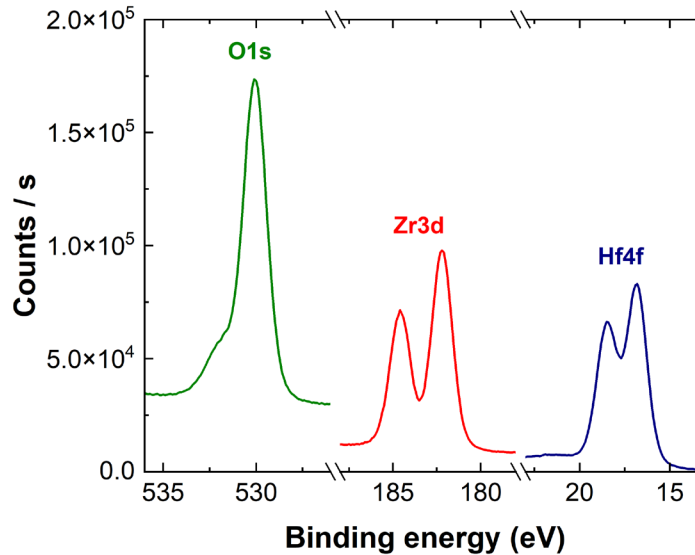
## Step 2: Composition - $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

2

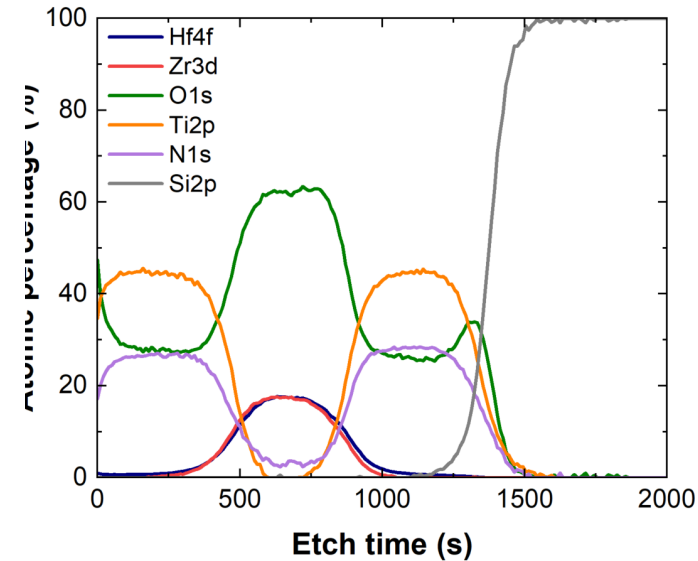
$\text{AB}_x$



XPS scans

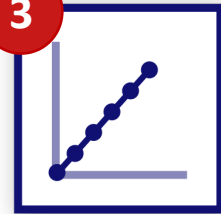


XPS depth profile (for  $\text{TiN}/\text{Hf}_x\text{Zr}_{1-x}\text{O}_2/\text{TiN}$  stack)



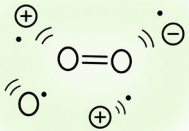
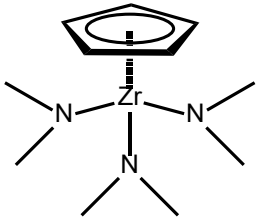
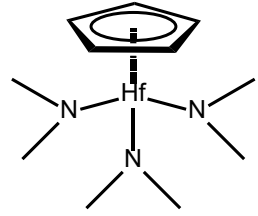
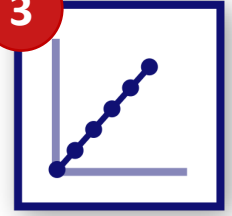
## Step 3: Thickness control – linear in #cycles?

- Verify thickness scales with cycle number
  - *In situ*: follow thickness during growth
  - *Ex situ*: prepare a thickness series
- Linear growth is necessary, but not proof of ALD
- Compare GPC with literature values, should match!
- Pay attention to the substrate temperature

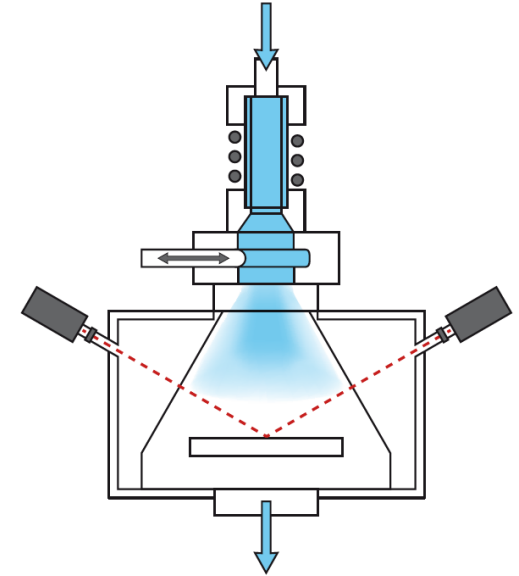
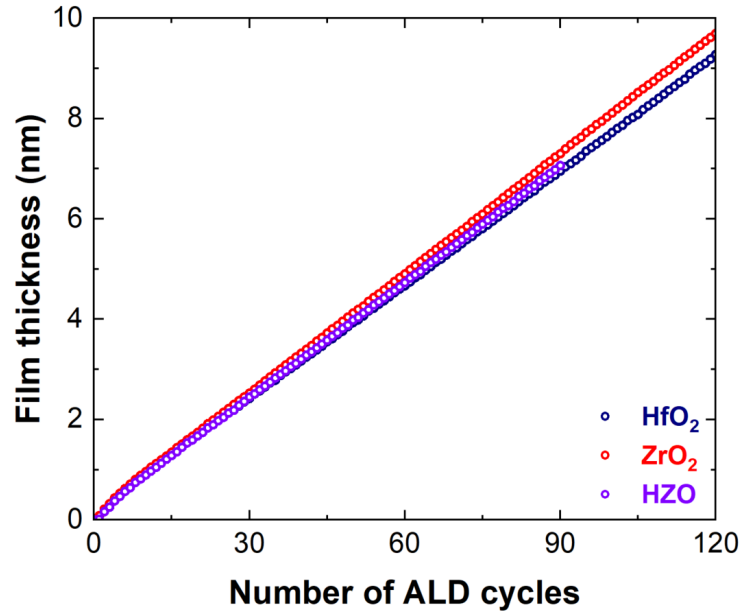


### Step 3: Thickness control – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

3



Thickness monitoring by *in situ* ellipsometry



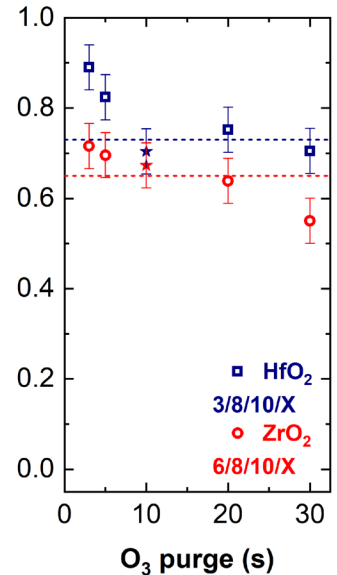
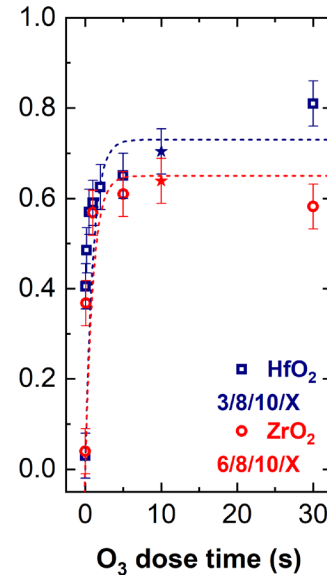
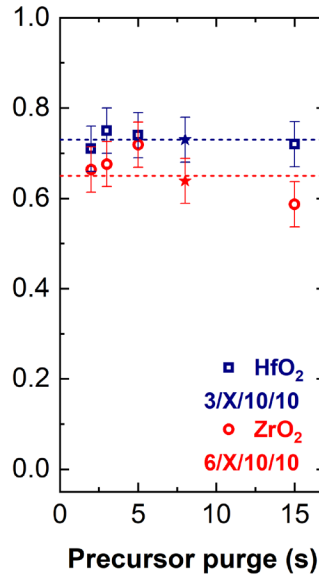
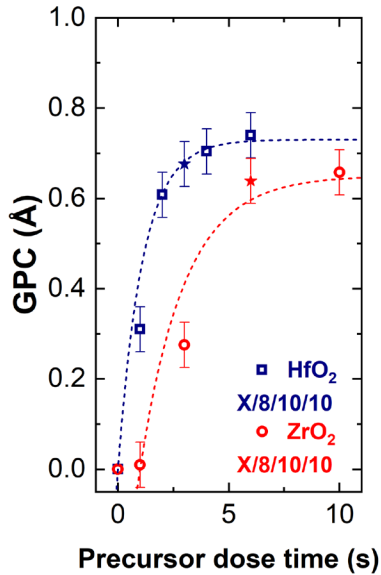
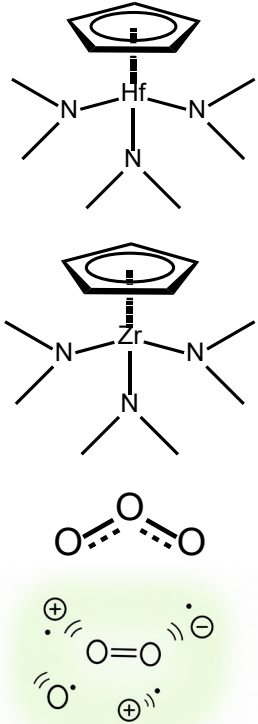
## Step 4: Saturation – Self-limiting growth?

- Start with sufficiently long dosing and purge times
- Vary one timing parameter at a time
- Look for a GPC plateau at long exposures
- Check zero-dose cases: no growth expected
- Confirm saturation for both half-cycles
- Repeat under “extreme” conditions if needed



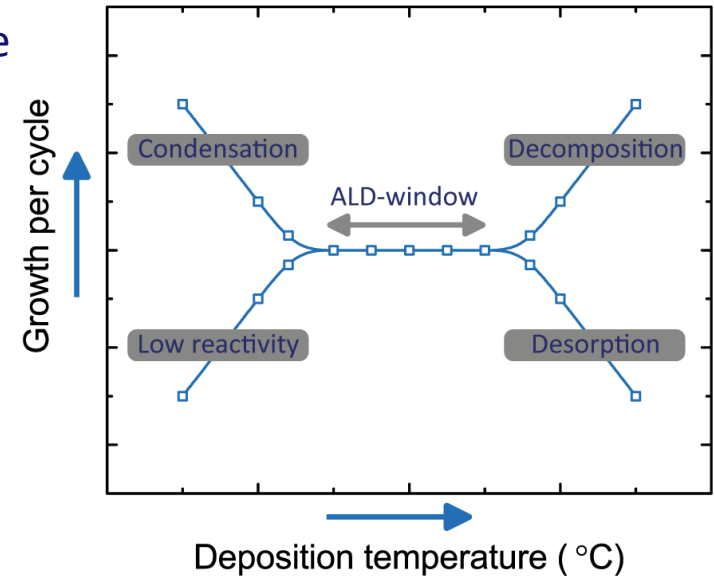
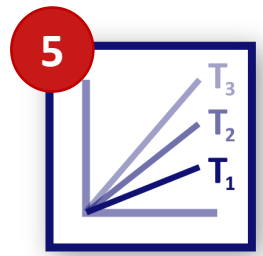
## Step 4: Saturation – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

4



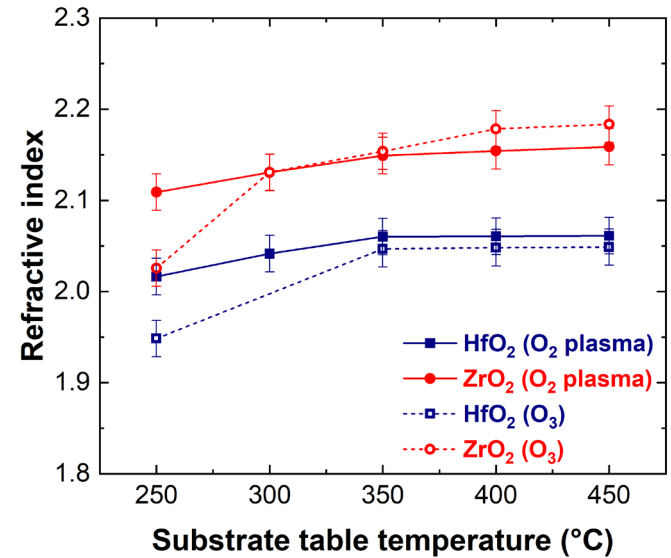
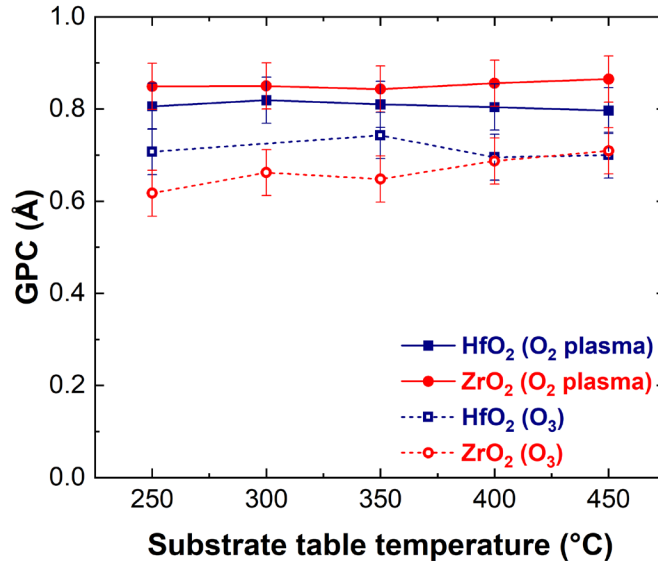
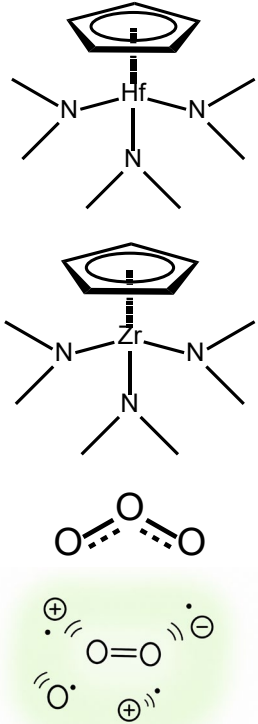
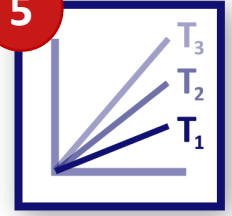
## Step 5: Temperature – Temperature window?

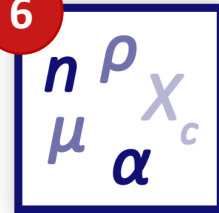
- Vary substrate temperature and measure GPC
- Check if self-limiting growth is maintained
- GPC does not need to be constant across the window
- Compare the temperature window with literature
- Check actual versus set substrate temperature



# Step 5: Temperature – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

5



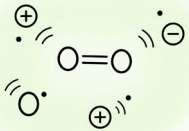
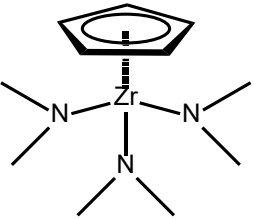
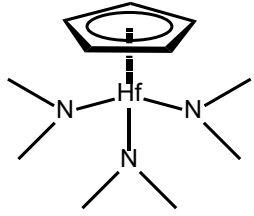
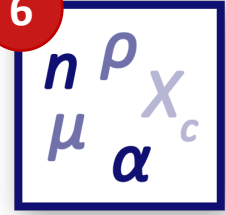


## Step 6: Film properties – Film quality?

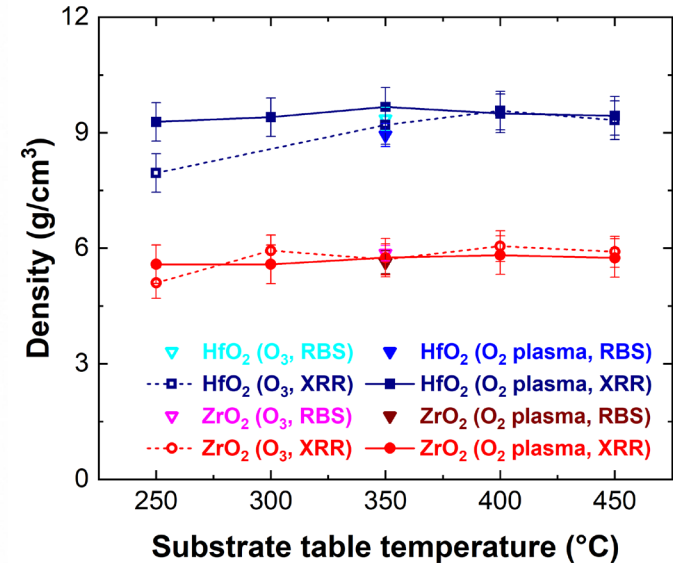
- Characterize the film beyond thickness
- Composition and impurities: XPS, RBS, SIMS, ...
- Structure and morphology: XRD, Raman, TEM, AFM, ...
- Density and optical properties: XRR, ellipsometry, ...
- Electrical properties: 4-point probe, Hall, devices, ...
- Verify temperature dependence
- Benchmark against literature

## Step 6: Film properties – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

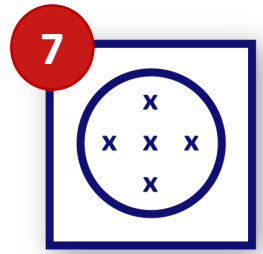
6



| Sample                  | M:O ratio       | H (at. %)      |
|-------------------------|-----------------|----------------|
| $\text{HfO}_2$ (ozone)  | $0.57 \pm 0.04$ | $6.2 \pm 0.5$  |
| $\text{ZrO}_2$ (ozone)  | $0.49 \pm 0.04$ | $6.6 \pm 0.5$  |
| $\text{HZO}$ (ozone)    | $0.50 \pm 0.04$ | $4.8 \pm 0.5$  |
| $\text{HfO}_2$ (plasma) | $0.60 \pm 0.04$ | $8.8 \pm 0.5$  |
| $\text{ZrO}_2$ (plasma) | $0.55 \pm 0.04$ | $7.8 \pm 0.5$  |
| $\text{HZO}$ (plasma)   | $0.53 \pm 0.04$ | $12.1 \pm 0.5$ |



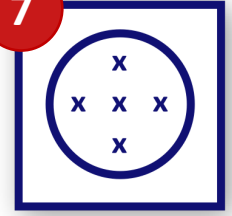
## Step 7: Uniformity – Same film across the substrate?



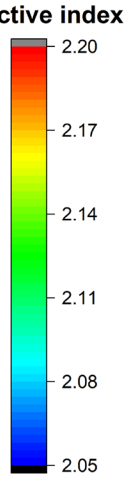
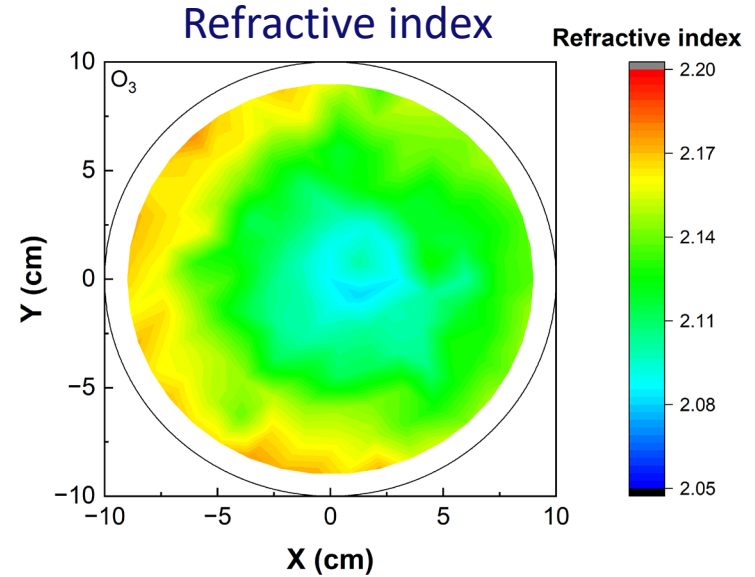
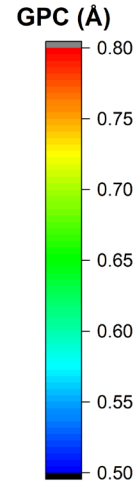
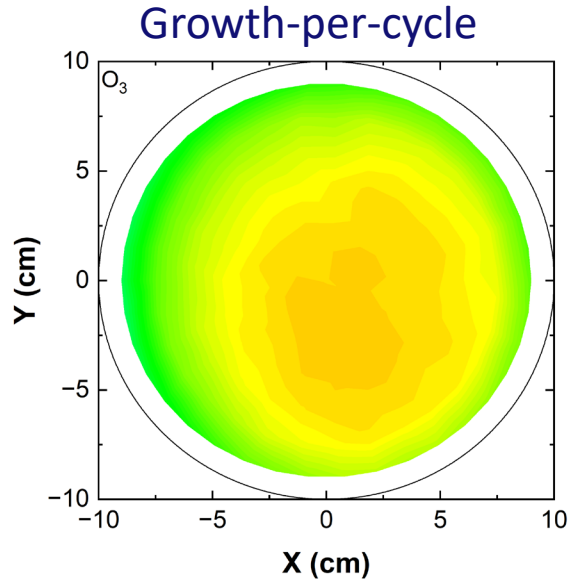
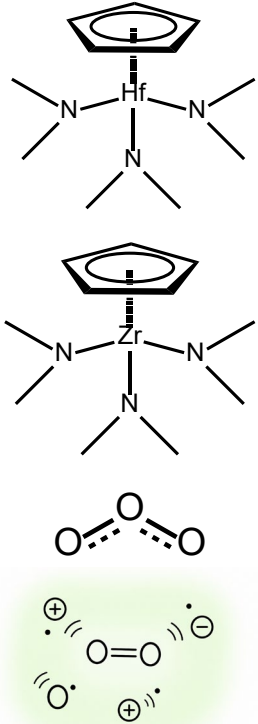
- Check uniformity on the largest substrate that fits
- Poor uniformity may indicate:
  - Incomplete saturation
  - Precursor decomposition
  - Gas-flow or reactor limitations
  - Influence of reaction products
  - Non-ideal ALD behavior
- Check both thickness and property uniformity

# Step 7: Uniformity – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

7



$\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$  with  $\text{O}_3$

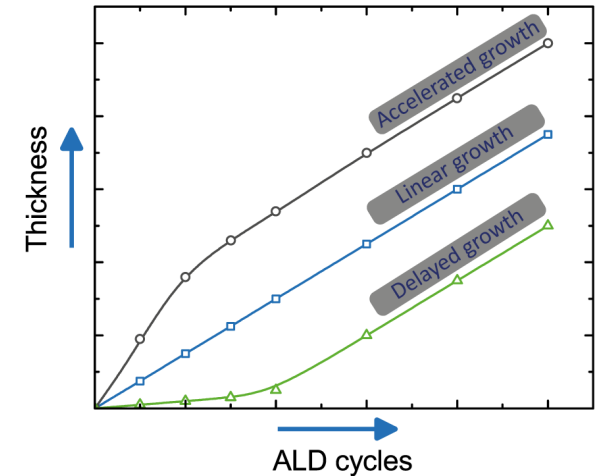


## Step 8: Nucleation – How does the film initially grow?

8

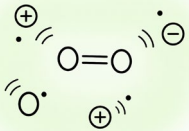
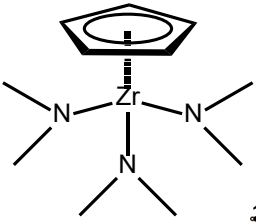
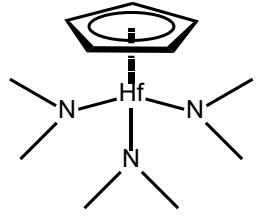


- Check growth versus cycle number on relevant substrates
- Pay attention to surface pretreatment and temperature
- Test all relevant surfaces in the final application
- Assess nucleation, island growth, and film closure
- Use early-growth tools such as ellipsometry, AFM, TEM or XPS

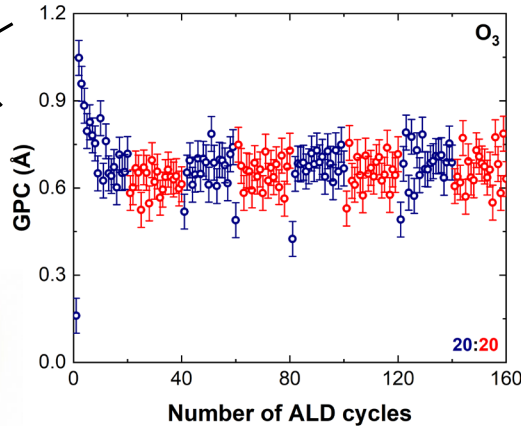


# Step 8: Nucleation – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

8

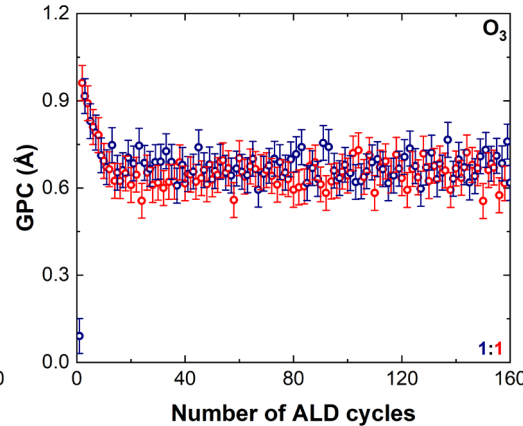


20 c.  $\text{HfO}_2$  : 20 c.  $\text{ZrO}_2$

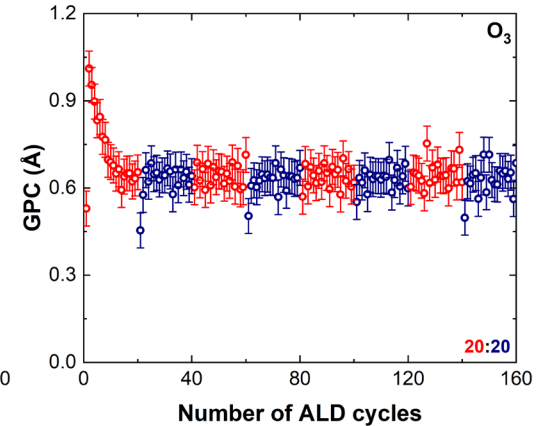


$\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$  with  $\text{O}_3$

1 c.  $\text{HfO}_2$  : 1 c.  $\text{ZrO}_2$

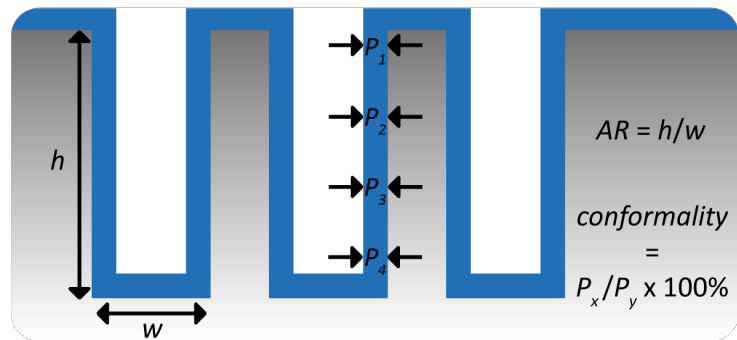


20 c.  $\text{ZrO}_2$  : 20 c.  $\text{HfO}_2$

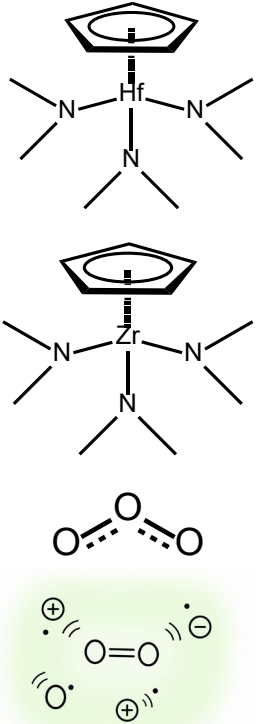


## Step 9: Conformality – Same film in 3D substrates?

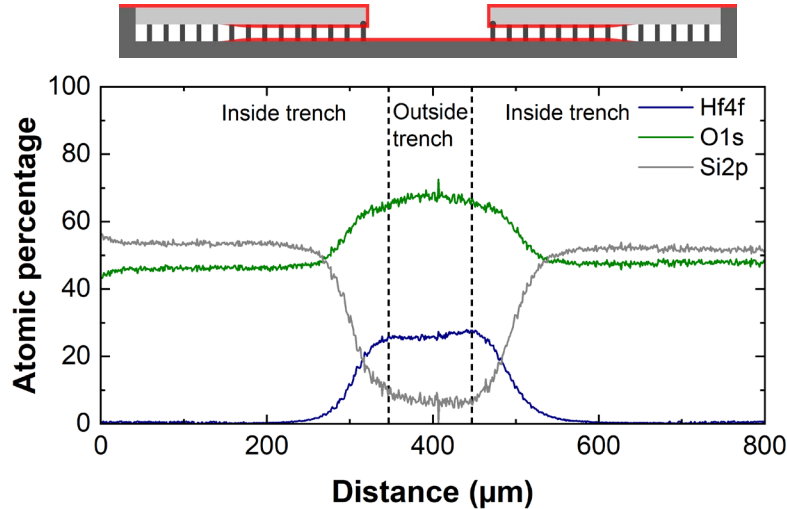
- Test conformality in relevant 3D structures
  - Trenches, pores, particles, test structures, ...
- Check thickness at different positions
- Check property conformality where possible too
  - Composition, etch rate, electrical behavior, ...
- Use structures with different aspect ratios
- Limited conformality can reveal transport limits or insufficient dosing



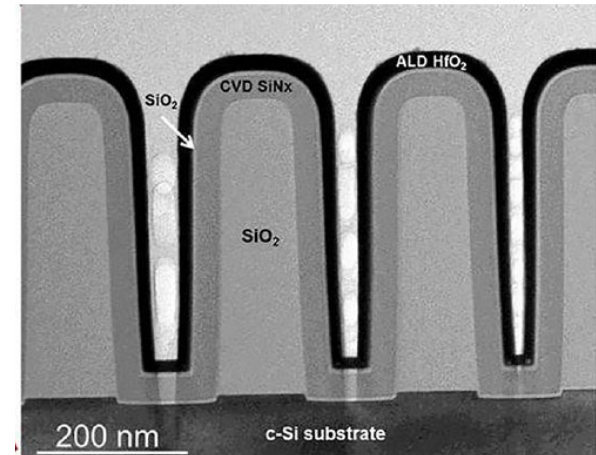
# Step 9: Conformality – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case



$\text{HfO}_2$  in Chipmetrics PillarHall structure



$\text{HfO}_2$  in etched trenches



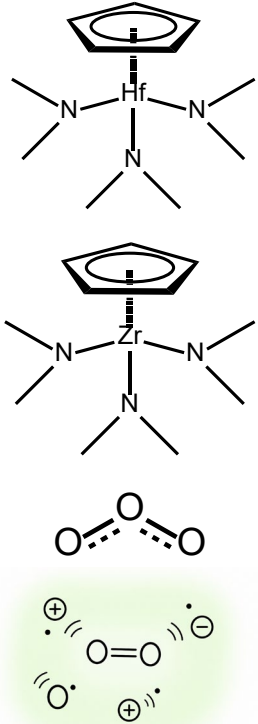


## Step 10: Other aspects – What else is important?

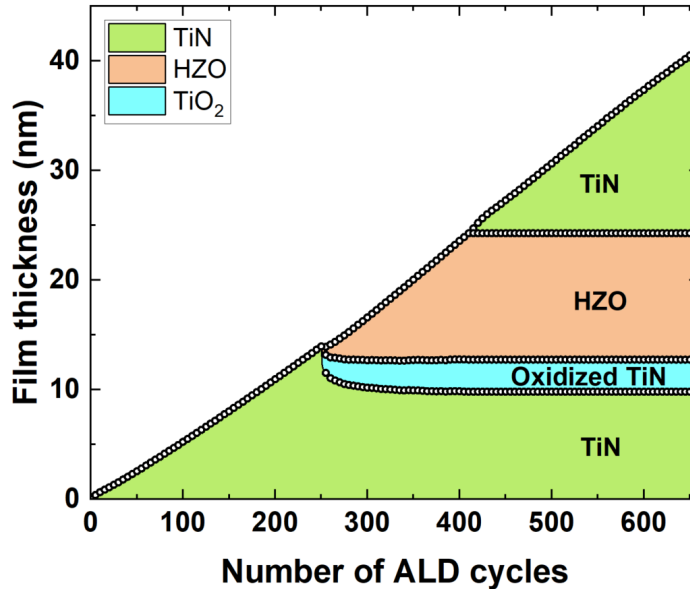
- Stability: check film changes over time and environment
- Reproducibility: repeat the same recipe many times
- Reactor history: consider wall conditioning and memory effects
- Precursor use: avoid excessive precursor consumption
- Integration: ensure compatibility with the process flow and device stack
- Throughput and scalability: match the needs of the application
- ...

# Step 10: Other aspects – $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ case

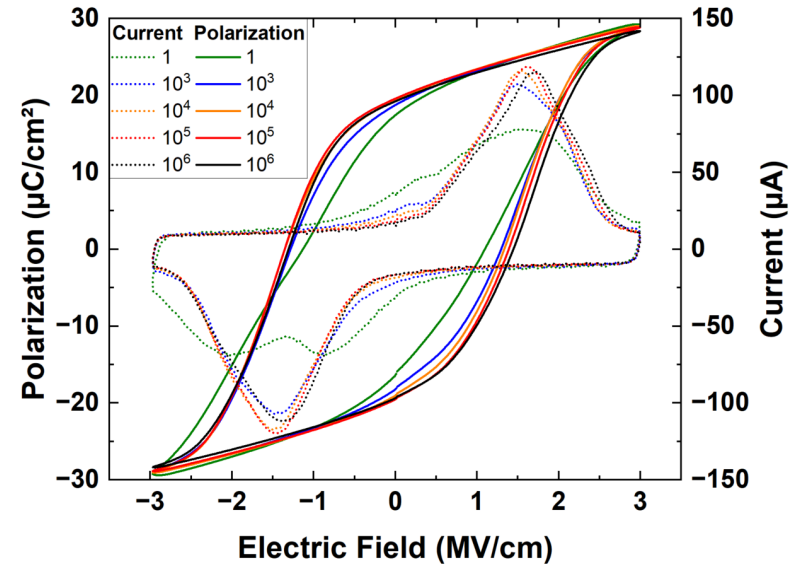
10



*In situ* ellipsometry during ALD of stack



Current and polarization

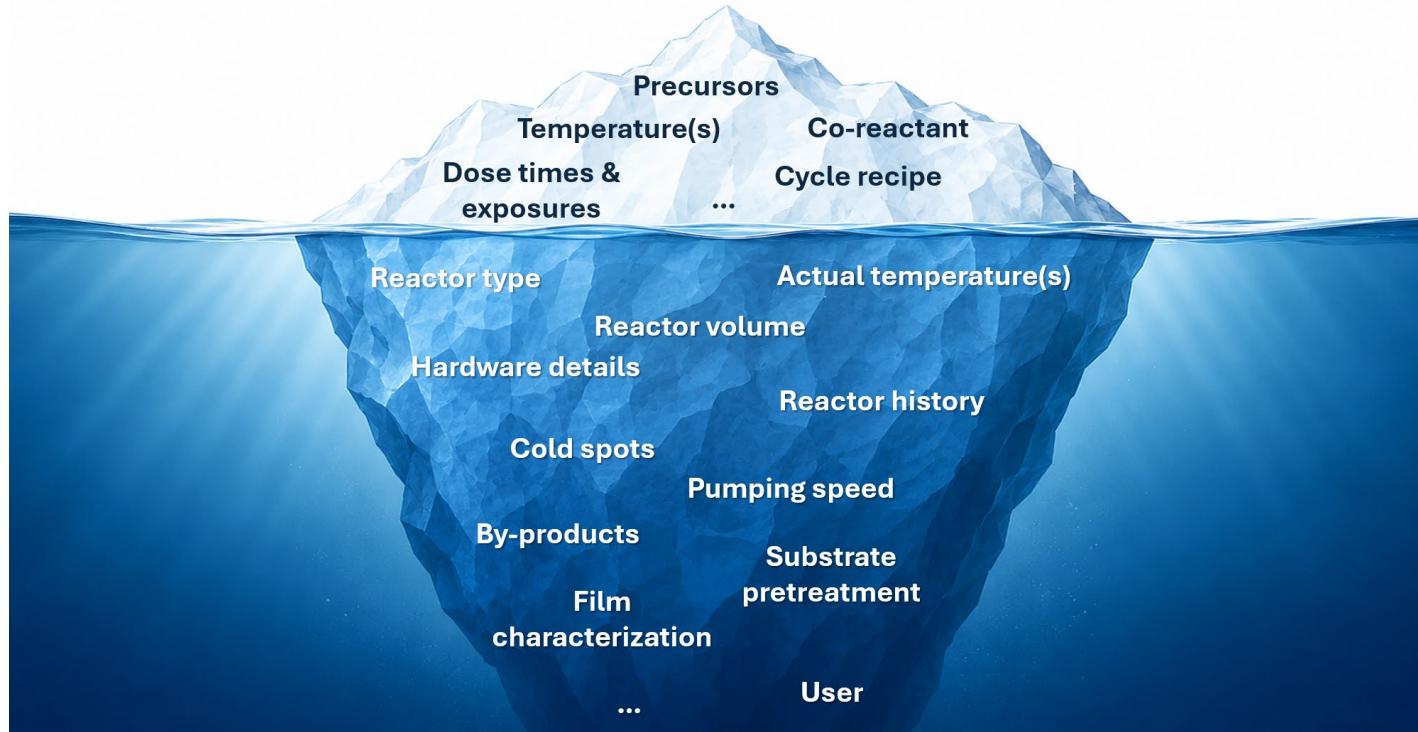


# ALD process development for $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$ – Illustrative example

| ALD development step | $\text{HfO}_2$ | $\text{ZrO}_2$ | $\text{Hf}_{1-x}\text{Zr}_x\text{O}_2$ |                                                                                      |
|----------------------|----------------|----------------|----------------------------------------|--------------------------------------------------------------------------------------|
| 1 Reactant selection | ●              | ●              | ●                                      |                                                                                      |
| 2 Composition        | ●              | ●              | ●                                      |                                                                                      |
| 3 Thickness control  | ●              | ●              | ●                                      | ← 0.70 Å/cycle for $\text{HfO}_2$<br>0.65 Å/cycle for $\text{ZrO}_2$                 |
| 4 Saturation         | ●              | ●              |                                        |                                                                                      |
| 5 Properties         | ●              | ●              | ●                                      | ← XPS, RBS, ellipsometry,<br>XRD, AFM, TEM<br>data available                         |
| 6 Temperature        | ●              | ●              | ●                                      |                                                                                      |
| 7 Uniformity         | ●              | ●              | ●                                      |                                                                                      |
| 8 Conformality       | ●              | ●              | ●                                      | ← Check conformality<br>Thickness/composition                                        |
| 9 Nucleation         | ●              | ●              | ●                                      |                                                                                      |
| 10 Other aspects     | ●              | ●              | ●                                      | ← $\text{TiN}/\text{Hf}_x\text{Zr}_{1-x}\text{O}_2/\text{TiN}$<br>Device performance |

# Challenge: the hidden complexity behind an ALD recipe

The recipe is only part of the process; many hidden factors can shape the outcome



# Outline

1. Quick introduction to ALD
2. ALD process development
3. **Towards AI-enabled ALD**
4. Challenges in frontier materials

# AI for ALD is emerging as a rapidly growing research theme

Literature spanning modeling, optimization, design, extraction, and agentic workflows\*

Christofides

|                    |                                                                                                                                  |                                       |
|--------------------|----------------------------------------------------------------------------------------------------------------------------------|---------------------------------------|
| Ding <i>et al.</i> | Machine learning-based modeling and operation for ALD of SiO <sub>2</sub> thin-films using data from a multiscale CFD simulation | Chem. Eng. Res. Des. 151, 131 (2019)  |
| Ding <i>et al.</i> | Machine learning-based modeling and operation of plasma-enhanced atomic layer deposition of hafnium oxide thin films             | Comput. Chem. Eng. 144, 107148 (2021) |
| Tom <i>et al.</i>  | Machine learning-based run-to-run control of a spatial thermal atomic layer etching reactor                                      | Comput. Chem. Eng. 168, 108044 (2022) |
| Tom <i>et al.</i>  | Machine Learning Modeling and Run-to-Run Control of an Area-Selective Atomic Layer Deposition Spatial Reactor                    | Coatings 14, 38 (2024)                |

Yanguas-Gil & Elam

|                           |                                                                                                                                         |                                          |
|---------------------------|-----------------------------------------------------------------------------------------------------------------------------------------|------------------------------------------|
| Paulson <i>et al.</i>     | Intelligent Agents for the Optimization of Atomic Layer Deposition                                                                      | ACS Appl. Mater. Inter. 13, 17022 (2021) |
| Yanguas-Gil <i>et al.</i> | Machine learning and atomic layer deposition: Predicting saturation times from reactor growth profiles using artificial neural networks | J. Vac. Sci. Technol. A 40, 82408 (2022) |
| Yanguas-Gil <i>et al.</i> | Surrogate models to optimize plasma-assisted atomic layer deposition in high aspect ratio features                                      | Phys. Plasmas 32, 73507 (2025)           |
| Yanguas-Gil <i>et al.</i> | Design and performance of AI agents interfacing with an atomic layer deposition tool                                                    | Rev. Sci. Instrum. 97, 053903 (2026)     |

Banerjee

|                           |                                                                                                                                      |                                          |
|---------------------------|--------------------------------------------------------------------------------------------------------------------------------------|------------------------------------------|
| Arunachalam <i>et al.</i> | Machine learning approach to thickness prediction from in situ spectroscopic ellipsometry data for atomic layer deposition processes | J. Vac. Sci. Technol. A 40, 12405 (2022) |
|---------------------------|--------------------------------------------------------------------------------------------------------------------------------------|------------------------------------------|

Shong

|                   |                                                                                                                                   |                             |
|-------------------|-----------------------------------------------------------------------------------------------------------------------------------|-----------------------------|
| Van <i>et al.</i> | Machine learning-based exploration of molecular design descriptors for area-selective atomic layer deposition (AS-ALD) precursors | J. Mod. Model 30, 10 (2024) |
|-------------------|-----------------------------------------------------------------------------------------------------------------------------------|-----------------------------|

Takoudis

|                           |                                                                                                                                   |                                          |
|---------------------------|-----------------------------------------------------------------------------------------------------------------------------------|------------------------------------------|
| Dogan <i>et al.</i>       | Bayesian Machine Learning for Efficient Minimization of Defects in ALD Passivation Layers                                         | ACS Appl. Mater. Inter. 13, 54503 (2021) |
| Kim <i>et al.</i>         | Machine learning-based analysis of the physio-chemical properties for the predictive thickness control of atomic layer deposition | IFAC-PapersOnLine 55, 626 (2022)         |
| Chiu <i>et al.</i>        | Ligand Engineering to Machine learning : Optimizing Ru ALD for Ultrathin Film Deposition                                          | Proc. ITC 2024                           |
| Häussermann <i>et al.</i> | Leveraging Bayesian Optimization Software for Atomic Layer Deposition: Single-Objective Optimization of TiO <sub>2</sub> Layers   | Materials 17, 5019 (2024)                |
| Seo <i>et al.</i>         | Deriving optimal atomic layer deposition process conditions using machine learning                                                | J. Ind. Inf. Integr. 47, 100879 (2025)   |
| Gu <i>et al.</i>          | AI-Driven Inverse Design of Complex Oxide Thin Films for Semiconductor Devices: A Case Study of Hf-Zr-O                           | Chem. Mater. 38, 5285 (2026)             |
| Navabi <i>et al.</i>      | Artificial intelligence in atomic layer deposition                                                                                | J. Vac. Sci. Technol. A 43, 60801 (2025) |

Bent & Tonner-Zech

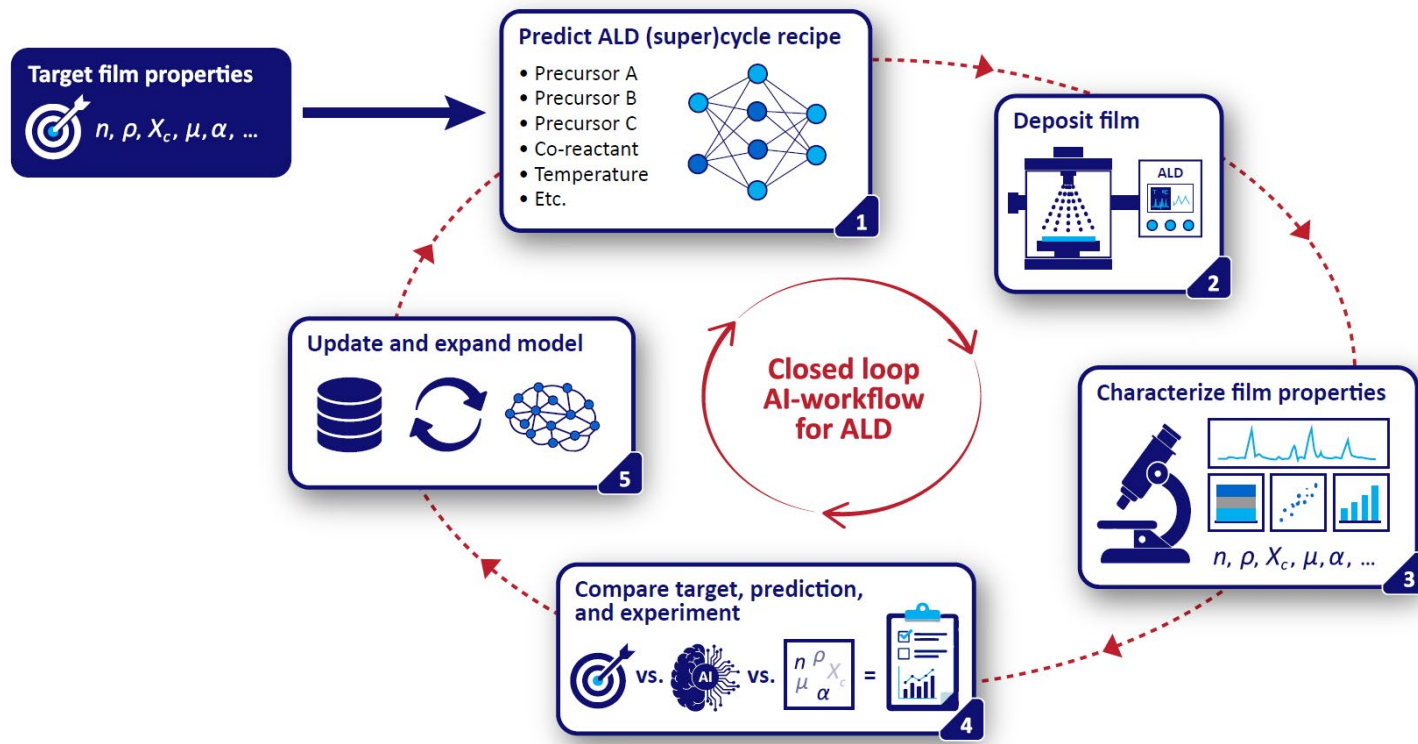
|                            |                                                                                                                                                                                       |                                         |
|----------------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|-----------------------------------------|
| Ramachandran <i>et al.</i> | Advancing atomic layer deposition through artificial intelligence: Generative modeling and predictive analytics for TiO <sub>2</sub> growth rate prediction from TDMAT/O <sub>3</sub> | Surf. Coat. Technol. 526, 133341 (2026) |
| Westermayr <i>et al.</i>   | Opportunities and Challenges of Machine Learning in Atomic Layer Deposition                                                                                                           | Chem. Mater. 38, 2142 (2026)            |

D'Souza, Karasulu Mackus & Kessels

|                         |                                                                                                                                                      |                                          |
|-------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------|------------------------------------------|
| Sadruddin <i>et al.</i> | Designing an agentic AI workflow for structured information extraction from scientific text: A case study on atomic layer deposition of ZnO and IGZO | J. Vac. Sci. Technol. A 44, 32413 (2026) |
| Sadruddin <i>et al.</i> | SCHEMA-MINERpro: Agentic AI for Ontology Grounding Over LLM-Discovered Scientific Schemas in a Human-in-the-Loop Workflow                            | Semantic Web 17, 1 (2026)                |
| D'Souza <i>et al.</i>   | Publishing FAIR and machine-actionable reviews in materials science: The case for symbolic knowledge in neuro-symbolic artificial intelligence       | J. Vac. Sci. Technol. A 44, 32408 (2026) |

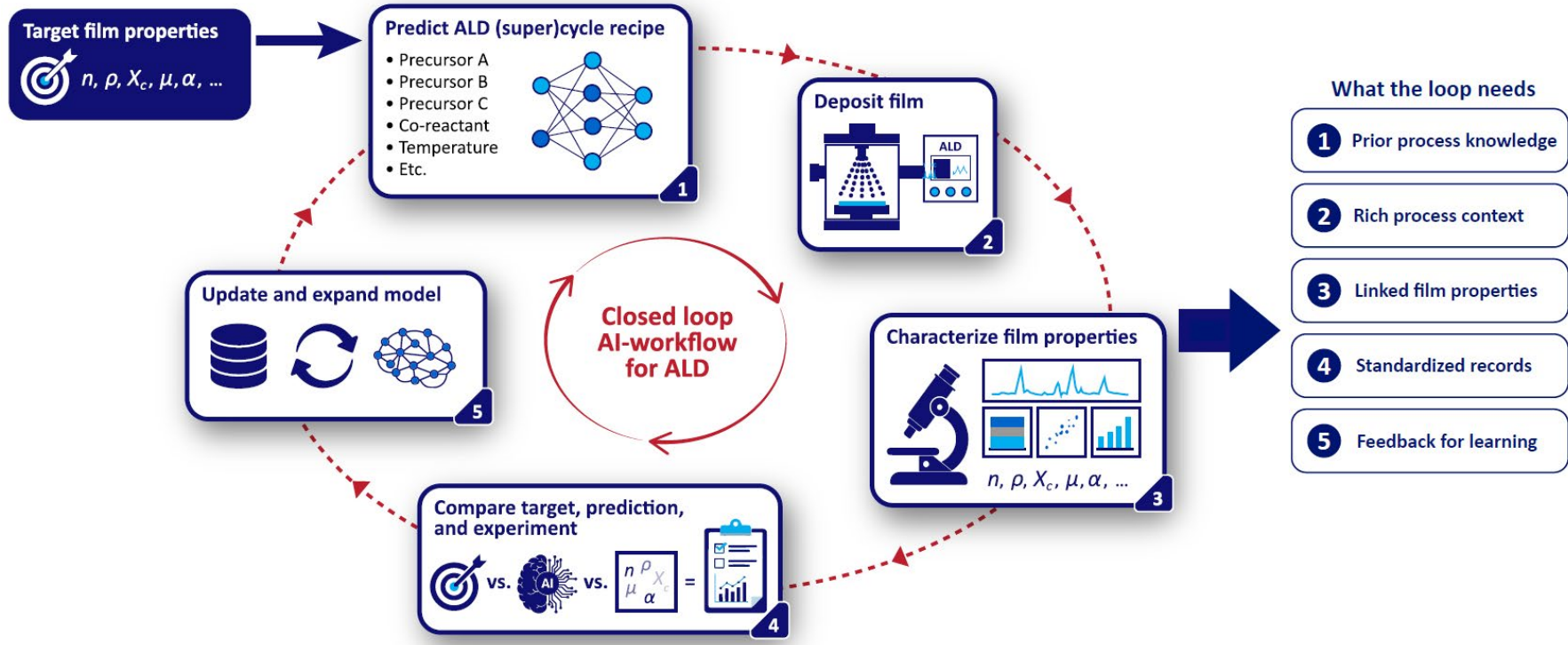
# A vision for AI-enabled ALD: closed-loop AI-workflow

From target properties to recipe prediction, experiments, and model learning



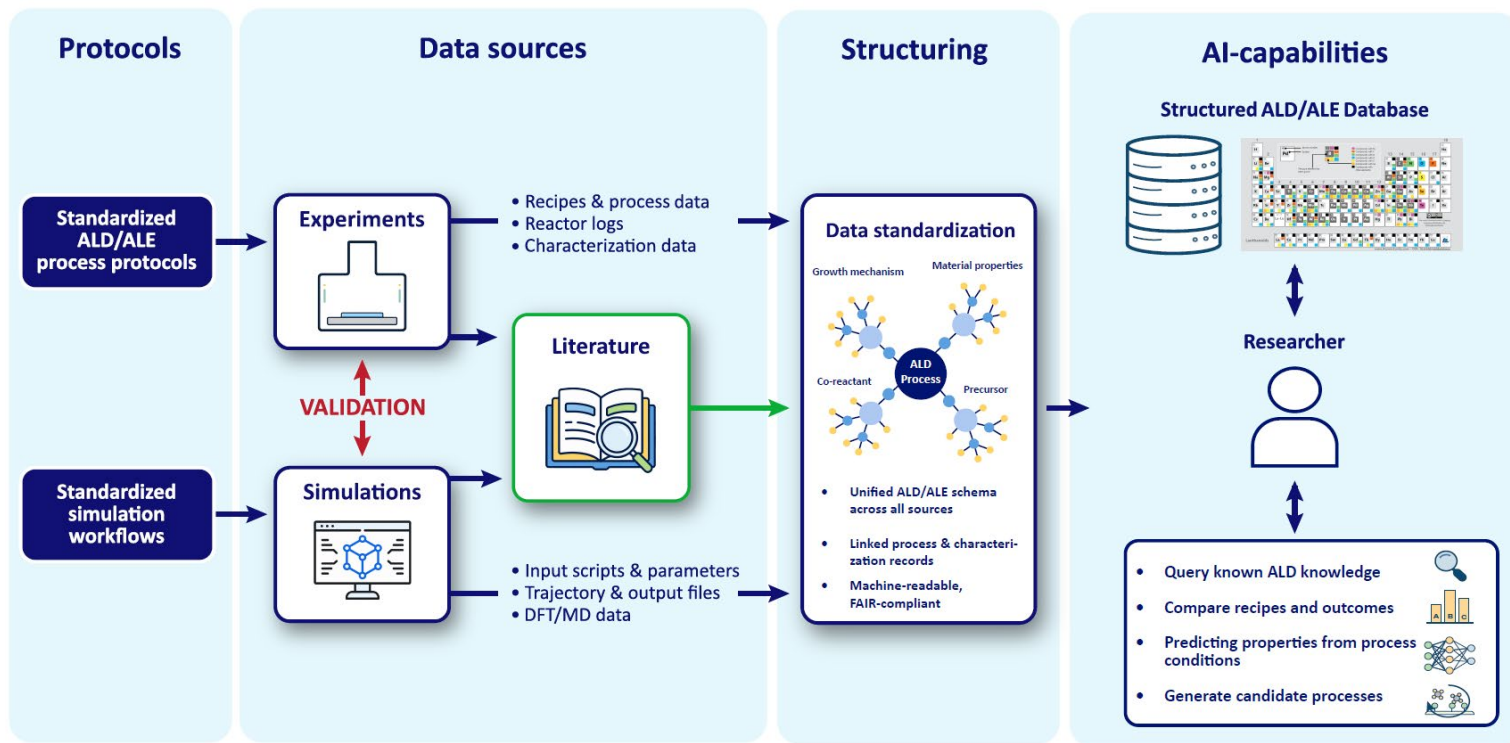
# What an AI-enabled ALD workflow needs from the data layer

The AI loop needs structured ALD knowledge, context, and feedback



# Making the ALD database AI-ready

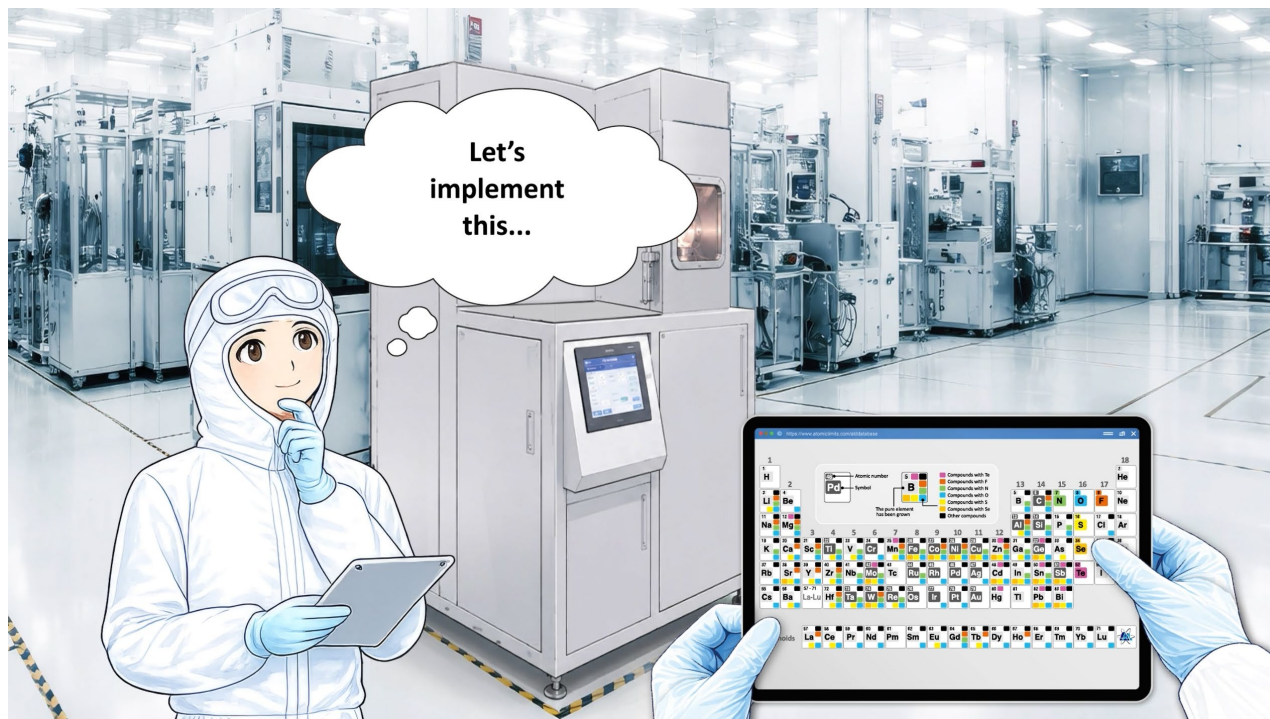
## Building the data layer needed for AI-enabled ALD



# Mock-up: an AI-ready ALD database in action

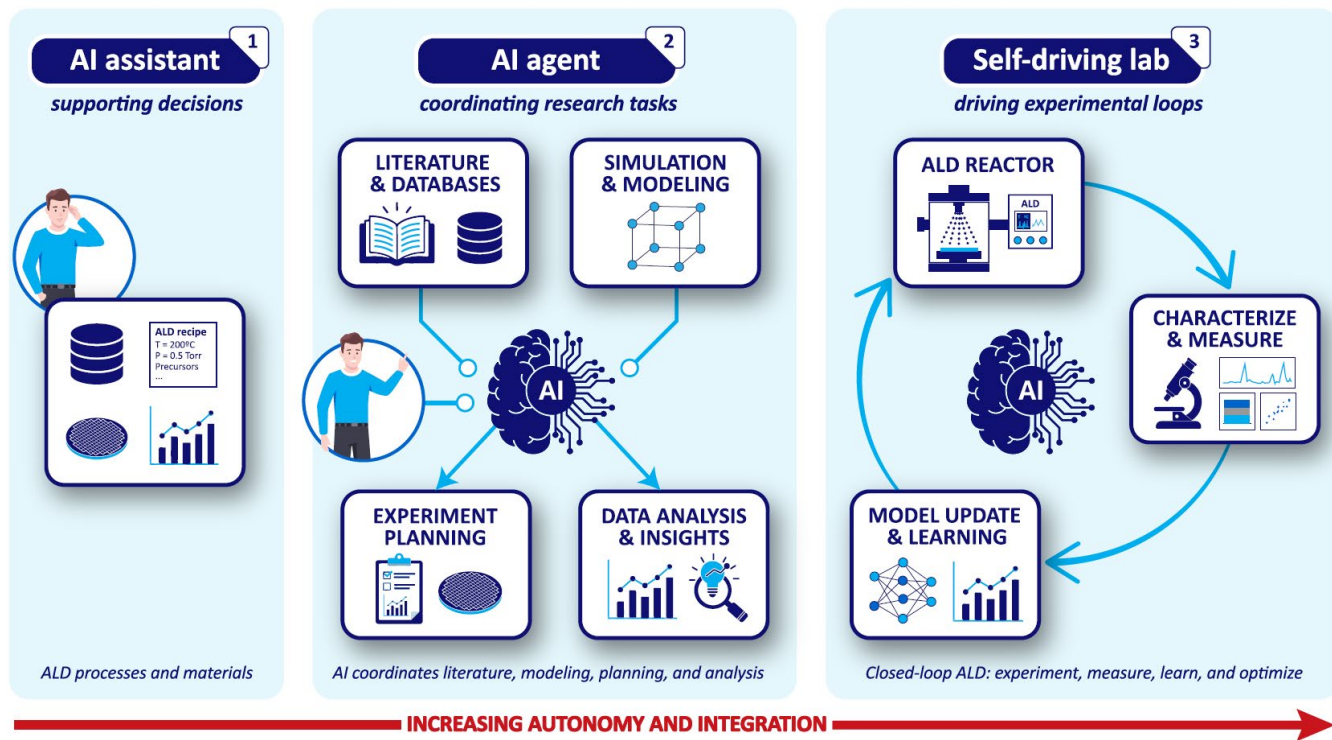
From today's ALD database toward a research tool for AI-enabled ALD

Video



# Outlook: from AI assistance to AI agents to self-driving labs

From supporting decisions, to coordinating research tasks, to driving experimental loops



# Outline

1. Quick introduction to ALD
2. ALD process development
3. Towards AI-enabled ALD
4. **Challenges in frontier materials**

# Materials with a difficult ALD chemistry

Representative material classes where robust self-limiting chemistry remains challenging

## Foundational Si- and C-based materials



Examples: a-Si, a-C, SiN<sub>x</sub>, SiCN, low-k, etc.

Limited thermal chemistries, often O<sub>3</sub> or plasma based (if any), conformality, low-temperature, ...

## Metals and their alloys



Examples: Cu, Co, Ni, Al, Ti, Ta, Nb, Au, Ag alloys, etc.

Limited thermal chemistries; reduction to metallic state, contamination, nucleation, film closure, ...

## Low-dimensional materials including 2D transition metal chalcogenides



Examples: MoS<sub>2</sub>, WSe<sub>2</sub>, h-BN, graphene, etc.

Stoichiometry, layer control, crystallinity, phase, ...

## Alkali-containing materials

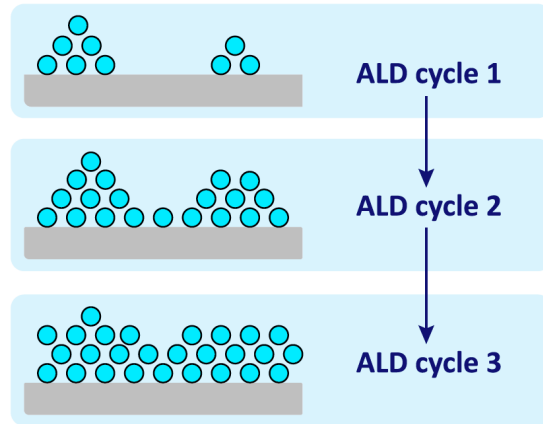


Examples: battery materials, Na/K-based films, etc.

Air-sensitive precursors, high mobility, hygroscopic films, and non-classical ALD chemistry...

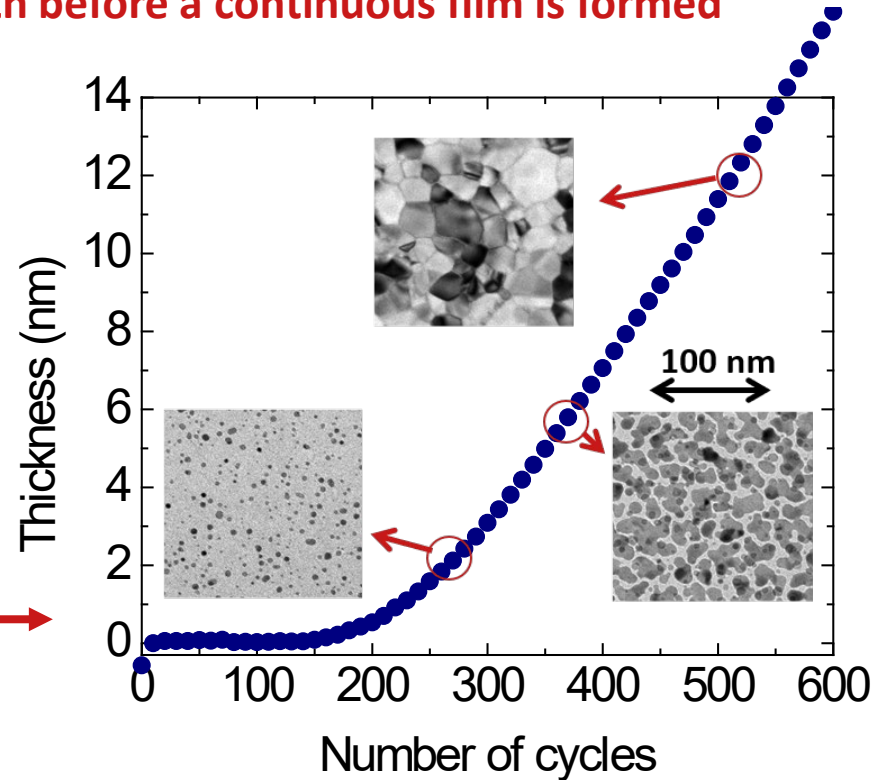
# Nucleation and film closure for ALD metal films

Metal ALD often starts with island growth before a continuous film is formed



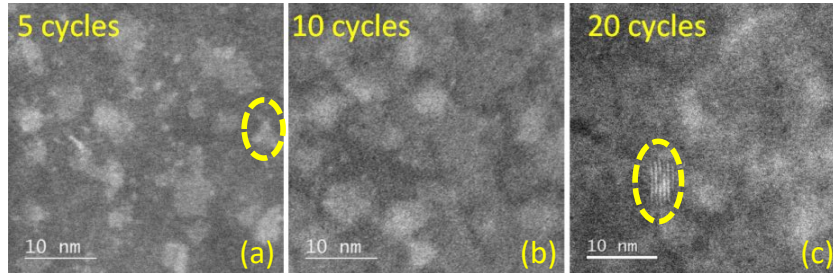
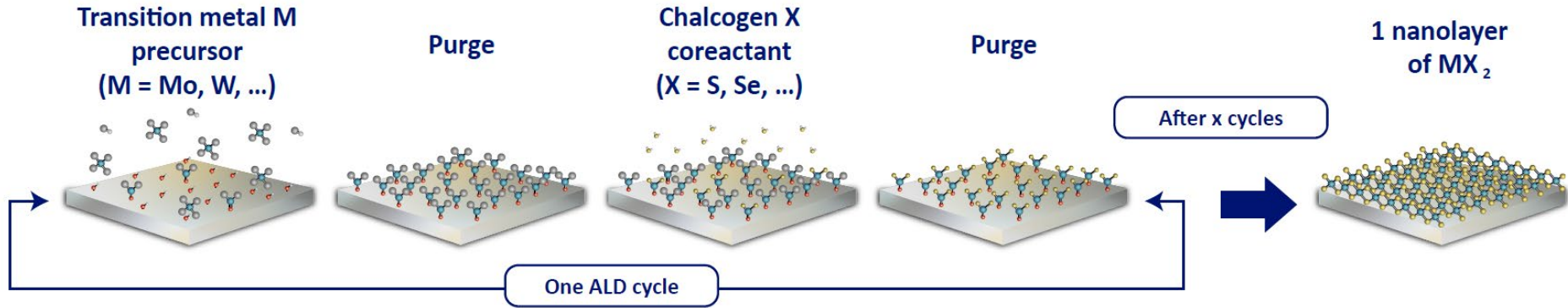
“Classic example”: Pt ALD on oxide surfaces

Also highly relevant for other Pt-group metals and coinage metals such as Ag and Au

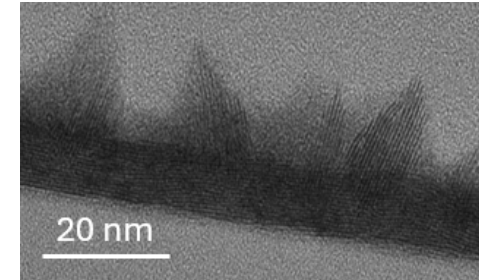


# Layer control and crystallinity for ALD 2D materials

Low-temperature & conformal — but large-grain, in-plane oriented layers seems hard



Grain size is limited to the tens-of-nm range



Out-of-plane-oriented structures

# ALD challenges in frontier materials

From making the material to achieving the required film functionality

## Preparing materials with difficult ALD chemistry

*Can the material itself be grown?*

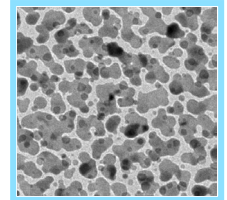
Intrinsic chemistry challenges:  
precursors · reactants · impurities



## Forming closed films, especially at ultralow thickness

*Can a continuous layer form early enough?*

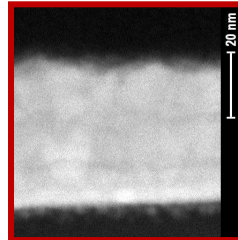
Nucleation challenges:  
nucleation · coalescence · closure



## Controlling elemental mixing beyond binaries

*Can elements be mixed at the atomic scale?*

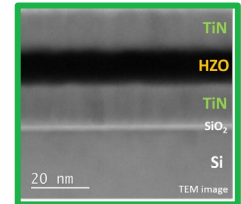
Process route challenges:  
mixing · interfaces · phase purity



## Achieving highly crystalline films

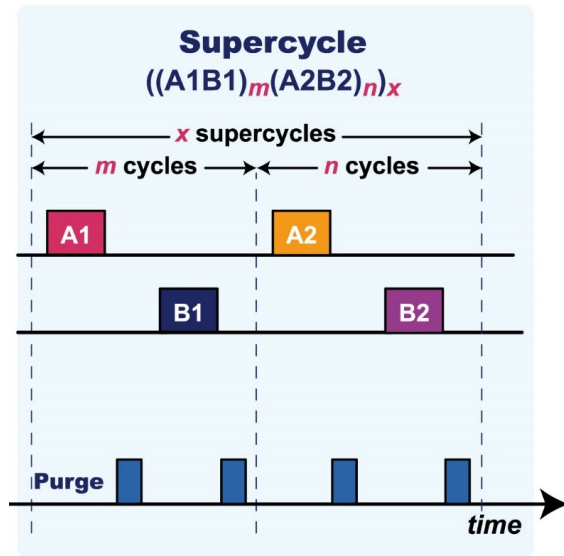
*Can a process deliver the right phase?*

Process route challenges:  
phase · orientation · grains

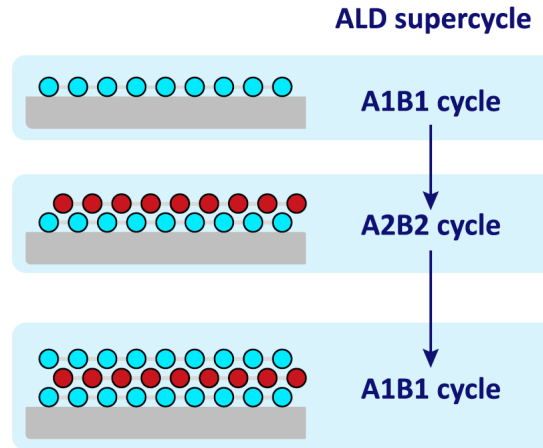


# ALD challenge: elemental mixing beyond binary compounds

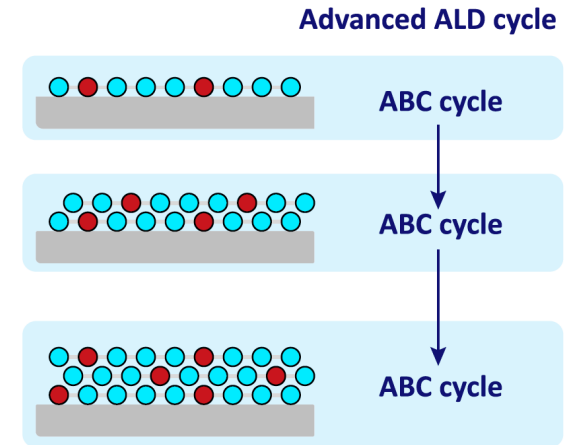
Average composition is not enough: the atoms also need to be mixed in the right way



Supercycles can result in nanolaminate-like films

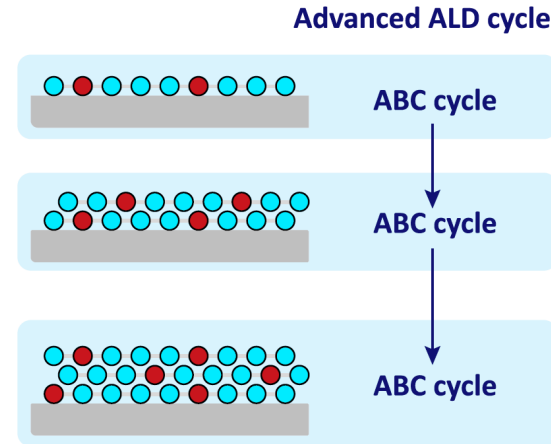
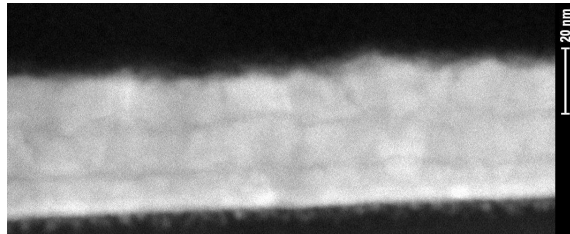
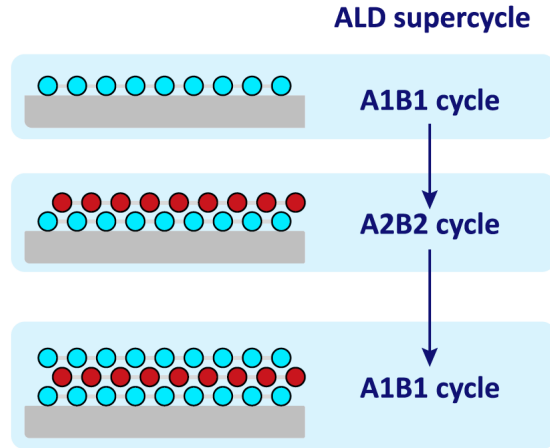


Advanced (super)cycles can promote more homogeneous mixing



# ALD of nanolaminate films vs. elemental mixing

Better distributed dopants or mixed elements through smart designed ALD processes



Approaches: Use of inhibitors, bulky precursors, modified sub-cycle sequence, co-dosing, atomic layer modulation, etc.

Kim *et al.*, J. Vac. Sci. Technol. A 28, 1111 (2010)

Yanguas-Gil *et al.*, Chem. Mater. 23, 4295 (2011)

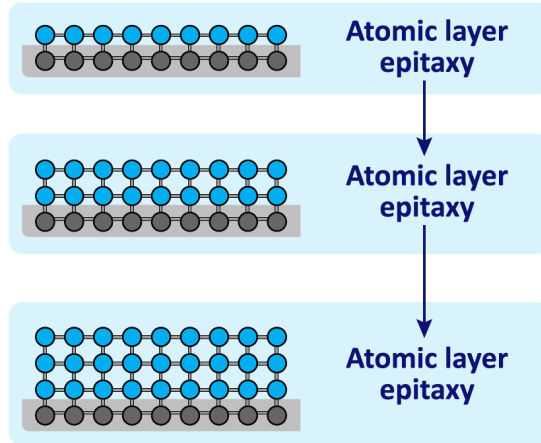
Wu *et al.*, Chem. Mater. 25, 4619 (2013)

Trinh *et al.*, ACS Nano 20, 17273 (2026)

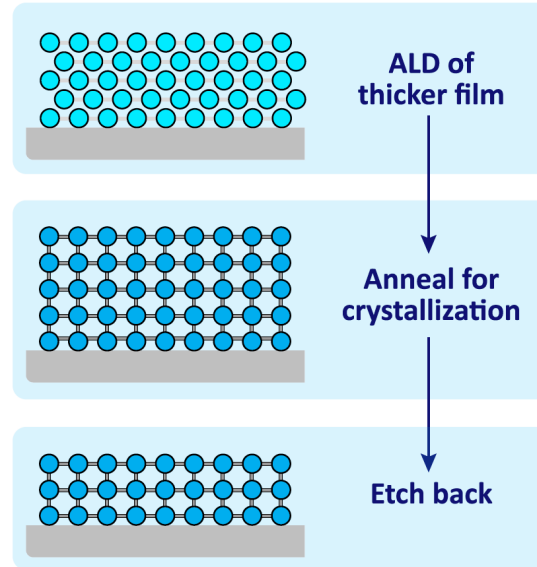
# ALD challenge: routes toward ultrathin crystalline films

Sometimes the solution is not a better recipe, but a smarter process route

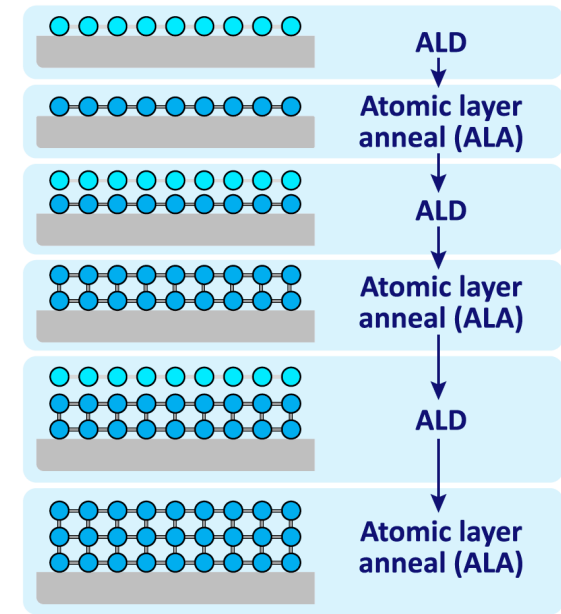
## Template-assisted crystalline growth



## Grow thicker → crystallize → etch back

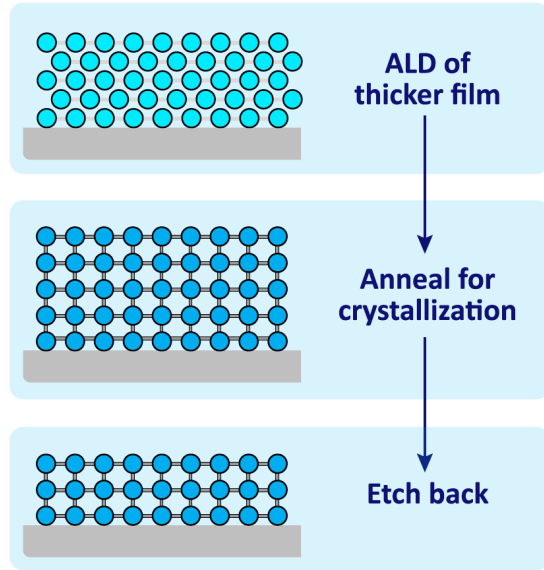


## Combining ALD with periodic activation/annealing

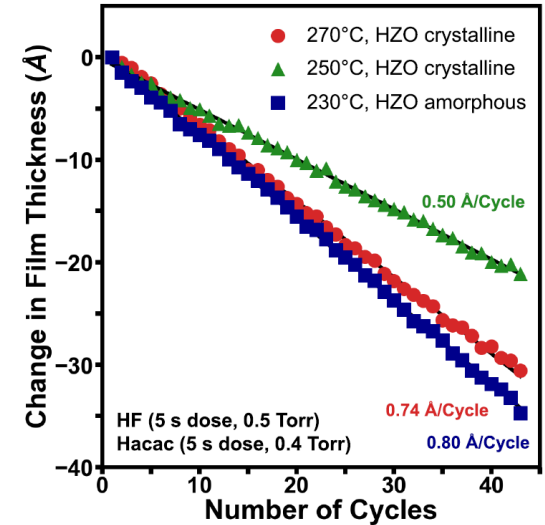
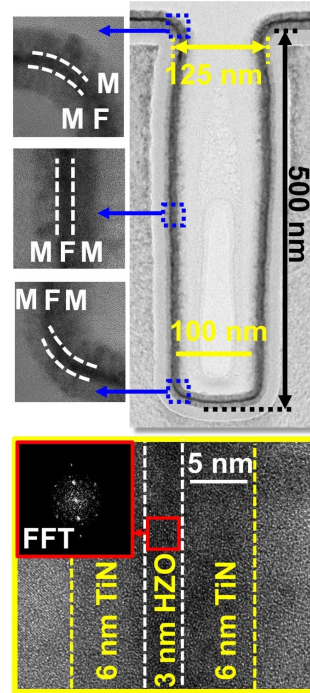


# ALD of ultrathin crystalline films using etch back

Combining ALD and ALE might be alternative for template-assisted crystalline growth



Example: ultrathin  $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$  (HZO) for 3D ferroelectric capacitors



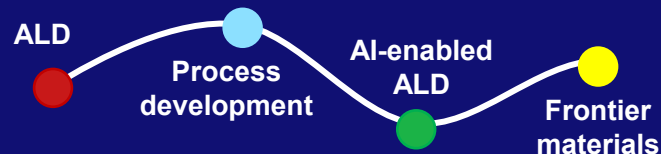
ALE of amorphous and crystalline  $\text{Hf}_x\text{Zr}_{1-x}\text{O}_2$

Feng *et al.*, VLSI symposium (2026)

Hoffmann *et al.*, Appl. Phys. Lett. 120, 122901 (2022)

Colleran *et al.*, Chem. Mater. 37, 5935 (2025)

# Connecting the story



- ALD is a powerful atomic-scale processing method for materials synthesis
- Developing robust ALD processes (including equipment, precursors, etc.) remains a central challenge for the community.
- Structured workflows, protocols, and databases make process development more systematic
- These structured workflows and data are also essential for AI-enabled ALD
- AI can help us address remaining challenges in frontier materials — together with related approaches such as ALE and ASD

# Acknowledgments

## Plasma & Materials Processing (PMP) group

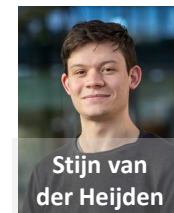


PMP principal investigators:

Adriana Creatore – Bart Macco – Adrie Mackus

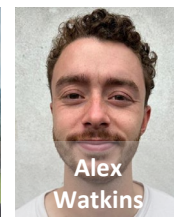
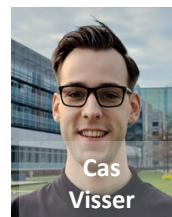
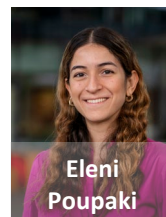
Julian Held – Ye Wang – Erwin Kessels

FluFe project:



Sponsor: 

AWASES project:







Sponsors:  