

# SCALABLE MANUFACTURING OF NANOSCALE MATERIALS AND INTERFACES

## *EDGE TO EXASCALE COMPUTING AND IN-SITU MEASUREMENTS*

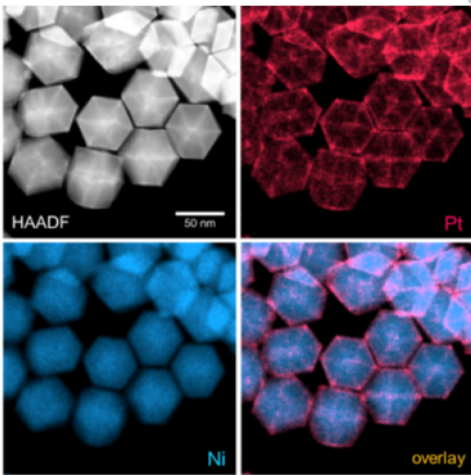
**SANTANU CHAUDHURI**

Professor of Materials Engineering, CME, University of Illinois at Chicago

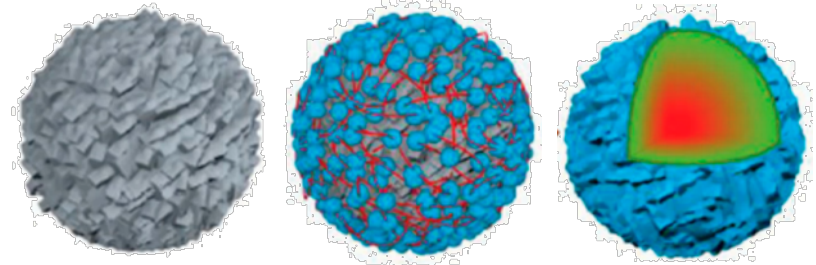
Director, Manufacturing Science and Engineering, ANL

# HIGH-PERFORMANCE ENERGY MATERIALS ARE OFTEN HARD TO SYNTHESIZE AND MANUFACTURE

Advanced materials are increasingly complex, requiring control of many synthesis steps



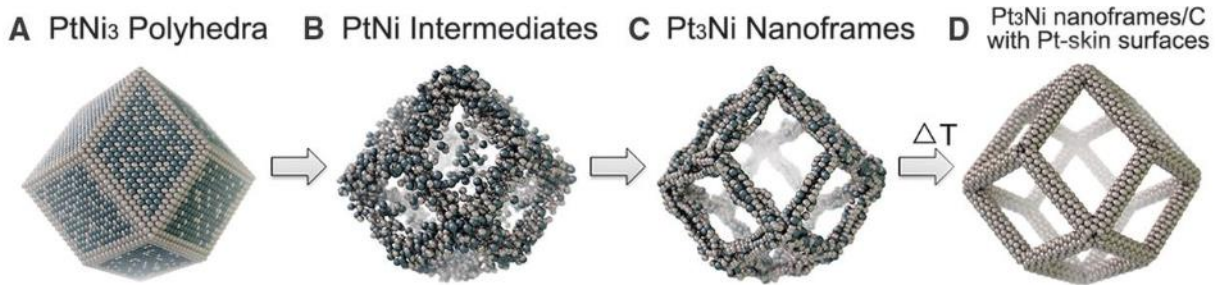
Pt-Ni nanoframe electrocatalysts for fuel cell applications: tightly controlled geometry and composition



Nickel-manganese-cobalt particles for cathode materials: multiple components, stratified

# ADVANCED ELECTROCATALYSTS: SYNTHESIS OF NANOFRAME CATALYSTS WITH HIGH ACTIVITY

C. Chen et al., *Science* 343, 1339–1343 (2014)



- Promising catalyst with potentially ground-breaking performance was discovered by an Argonne-Berkeley collaboration, through research funded by BES-MSED
- Gain in activity is attributed to the catalyst's complex three-dimensional geometry, which was created using a difficult, multi-step synthesis

**Initial synthetic methods were unable to produce this catalyst in sufficient quantity or uniformity to fully characterize its performance**

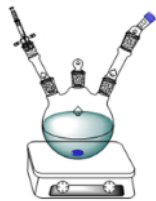
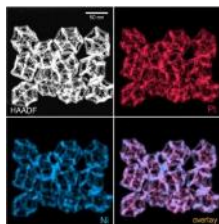
# ADVANCED ELECTROCATALYSTS: FROM DISCOVERY TO PROCESS R&D, WITH FEEDBACK TO SUPPORT DISCOVERY

Materials Engineering Research Facility (MERF)

## Discovery Science

### Material Discovery

- Poor batch-to-batch reproducibility
- Small quantities (mg) did not allow for full material characterization



### Innovative Batch Process

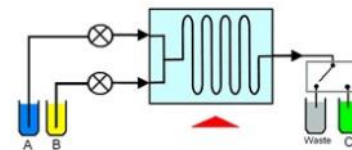
- Reproducible, gram-quantity samples
- Limited scalability
- Not suitable for manufacturing



Gram-quantity samples enabled full characterization of the new material

### Continuous-Flow Process

- Practically unlimited quantities
- Uniform quality of material
- **Suitable for manufacturing**



Continuous-flow reactor and instrumentation provide unique insights into synthetic mechanisms

# FLAME SPRAY PYROLYSIS: A VISION FOR MOVING FROM PREDICTIONS TO DIRECT MATERIALS SYNTHESIS

High-temperature synthesis (FSP) enables scalable synthesis of metastable materials

Materials Engineering Research Facility (MERF)

## Discovery Science

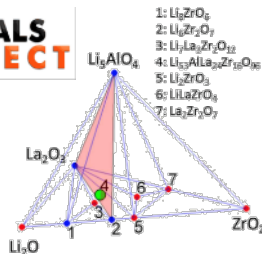
### Material Discovery: LLZO

- Al-doped LLZO (lithium lanthanum zirconium oxide) is a solid electrolyte
- Desired phase is metastable and requires careful synthesis



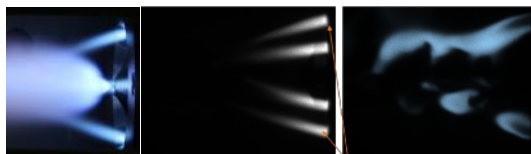
**MATERIALS PROJECT**

Materials Project is supported in part by BES



### Flame Spray Pyrolysis (FSP)

- In situ* diagnostics and *ex situ* analysis provide insights into complex phase behavior and gas-phase chemistry, enabling intelligent process manipulation



FSP with *in situ* diagnostics will aid understanding synthetic pathways and improve materials databases

### Scalable Synthesis and Validated Industry Interest

- Cabot uses FSP to manufacture relatively simple materials
- Now partnering with Argonne to advance this technology for complex battery materials

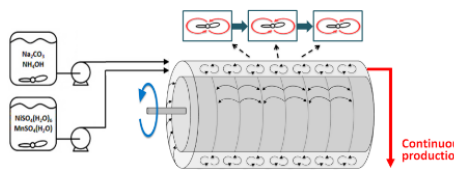
**CABOT**

Argonne  
NATIONAL LABORATORY

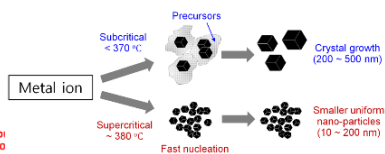
# ADVANCED MATERIALS MANUFACTURING RESEARCH

Expanding focus to advanced materials synthesis and processing

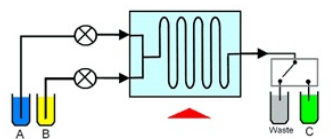
## Taylor Vortex



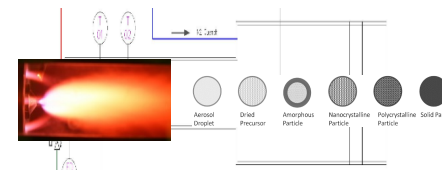
## Hydrothermal



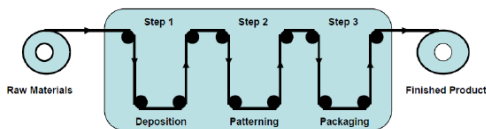
## Continuous Flow



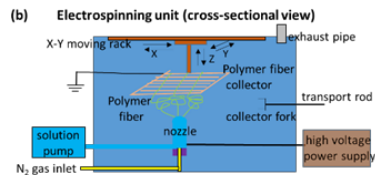
## Flame Spray



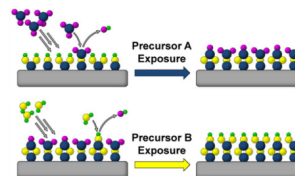
## Roll to Roll



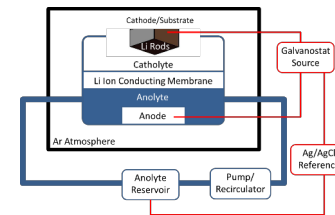
## Electrospinning



## Atomic Layer Deposition

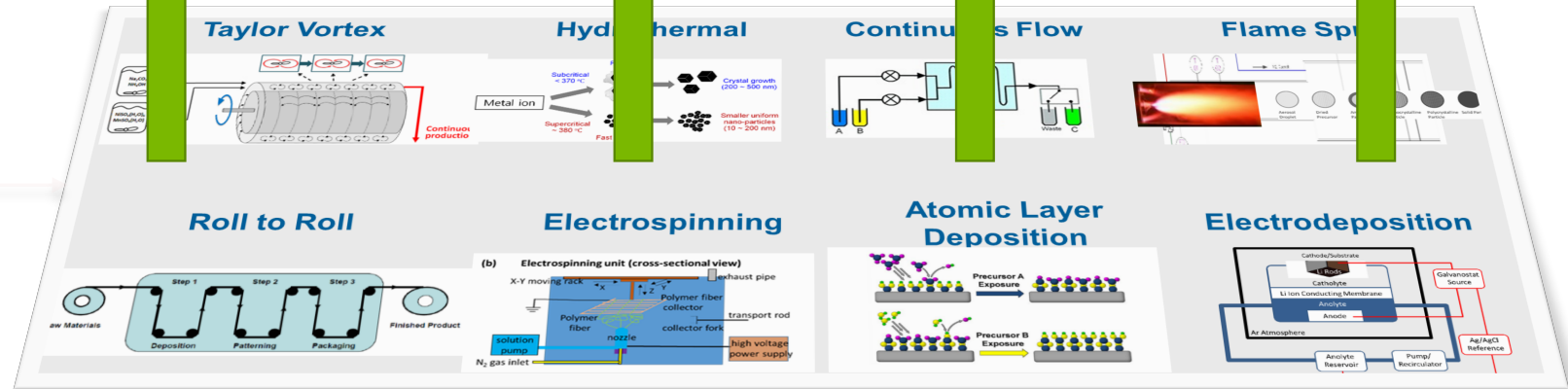


## Electrodeposition



# HPC for Modeling, Simulations, and Optimization

## Data Management and Analytics (AI/ML)



# ALCF 2021 EXASCALE SUPERCOMPUTER – A21

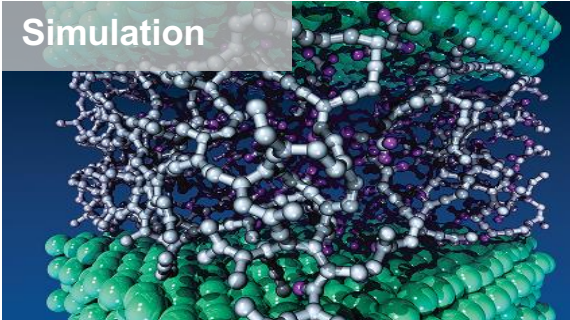
Aurora supercomputer planned for 2021

**Scaled up to over 1000 PF**

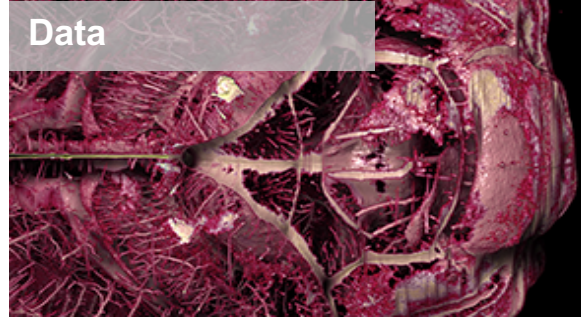


Support for three “pillars”

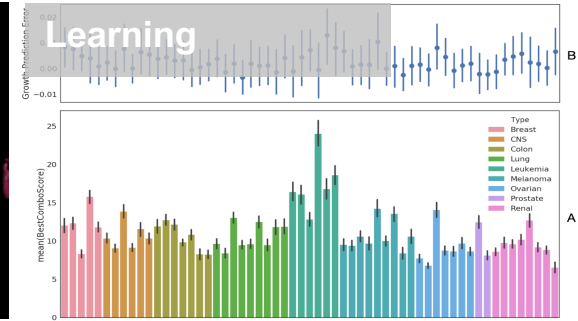
Simulation



Data



Learning





# MANUFACTURING SCIENCE AND ENGINEERING INITIATIVE

MERF + ALCF

Computing (AI, ML, HPC) for modeling, simulation, and learning

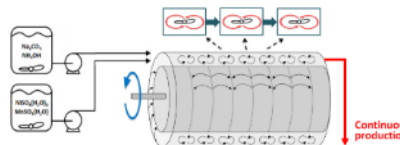
MERF + APS

Characterization (sensing, imaging, spectroscopy)

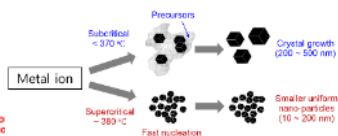
MERF

Advanced materials synthesis and processing technologies

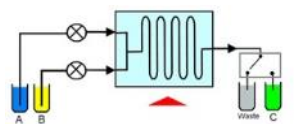
Taylor Vortex



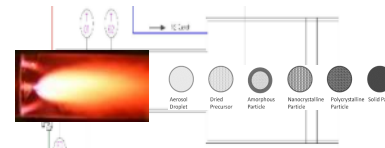
Hydrothermal



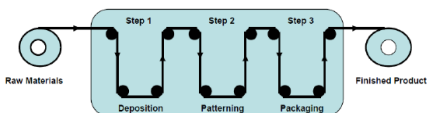
Continuous Flow



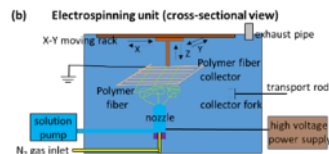
Flame Spray



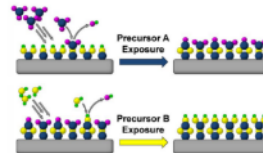
Roll to Roll



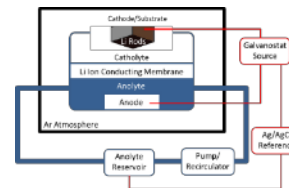
Electrospinning



Atomic Layer Deposition



Electrodeposition



# COLLABORATIONS

## Internal Cross-Laboratory and External with University and Industry

### Advanced Manufacturing: Related Workshops at Argonne

#### RECENT

- **Advanced Manufacturing Workshop**, ANL, August 31, 2017
  - Involved scientists and engineers across laboratory divisions
  - Explored advance manufacturing interest areas
- **High Performance Computing for Manufacturing Workshop**, ANL, Jan 2018
  - Brought together two communities: HPC and AM
  - Identified top areas of common interest: machine learning and multi-scale modeling
- **Additive Manufacturing Collaborative Workshop**, ANL, May 22, 2018
  - Organized by AMO, NNSA and ANL
  - Improved interactions with sponsors and expanded collaborations

#### UPCOMING

- **Machine Learning and Advanced Manufacturing Workshop**, ANL, Sep 2018
  - Organized by EGS and CLS
  - Breakout discussions of strategy and future project ideas
- **Artificial Intelligence Workshop**, ANL, November, 2018
  - Participation of DOE, Industry and Academia
  - Sessions dedicated to intelligent materials design and advanced manufacturing

### FACILITATING ENGAGEMENT WITH FOCUS UNIVERSITIES

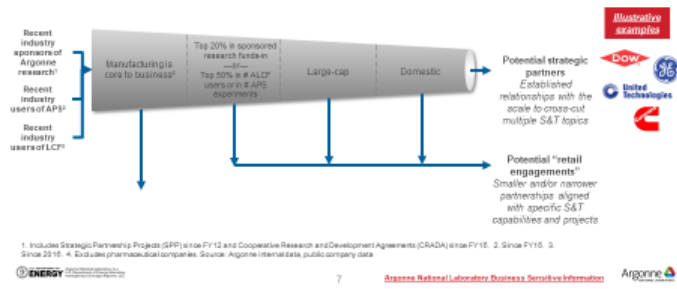
- Multiple visits between institutions (including UIC, UW-Madison, COD, etc)
- Fostering inter-institutional collaboration and agreements, which may lead to new institutes
- MSEI is part of new MOU with University of Wisconsin Madison's COE



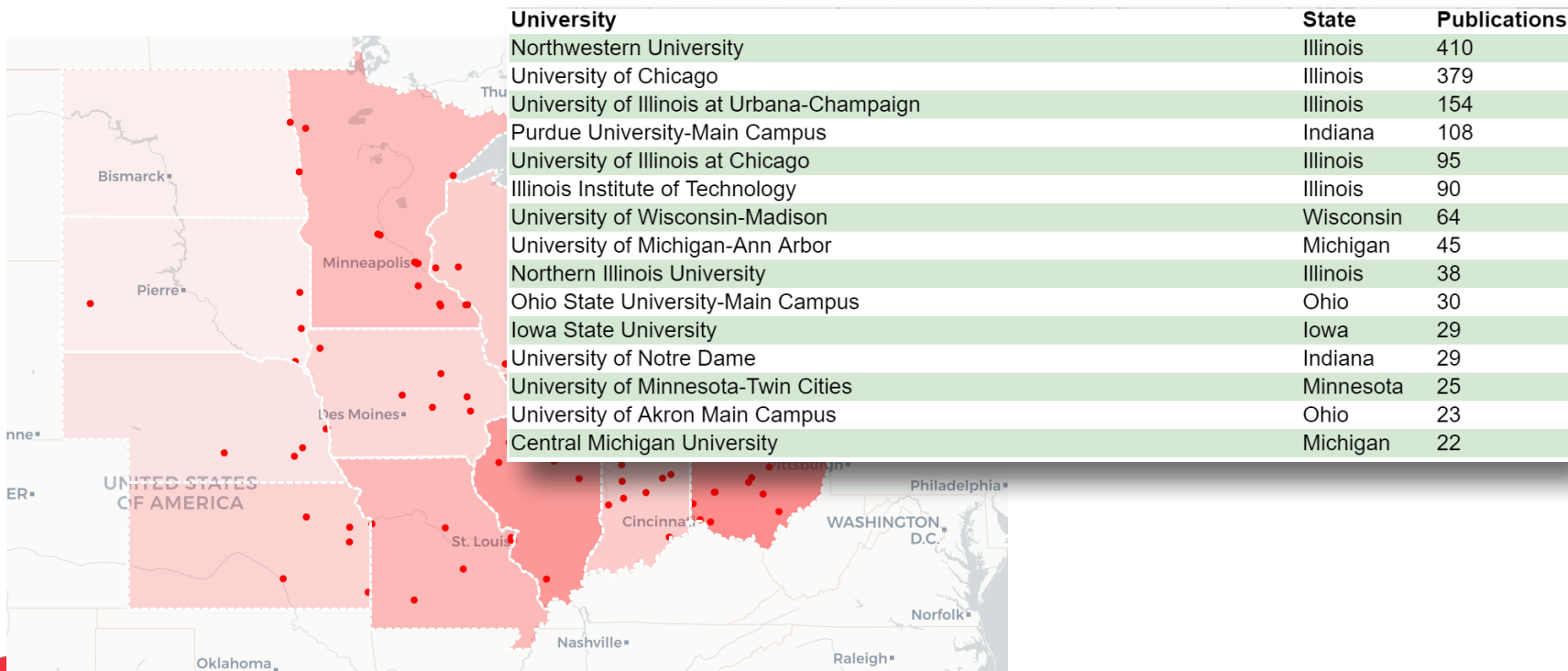
A visit by College of DuPage in April 2018 to their Innovation DuPage incubator

### USING DATA TO SORT AND MAP ARGONNE'S RELATIONSHIPS WITH MANUFACTURERS

And learn where Argonne has a "foot-in-the-door"

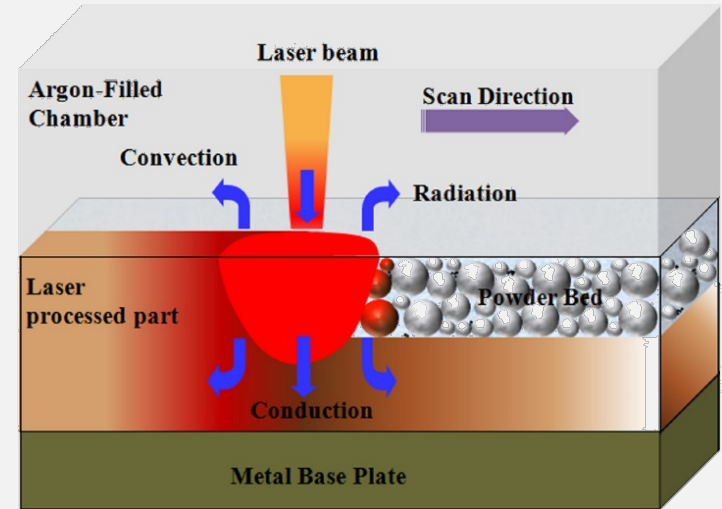


# MIDWEST PARTNERSHIP IN MANUFACTURING



# NANOSCALE CONTROL OF PROCESSING CONDITIONS FOR INTERFACES

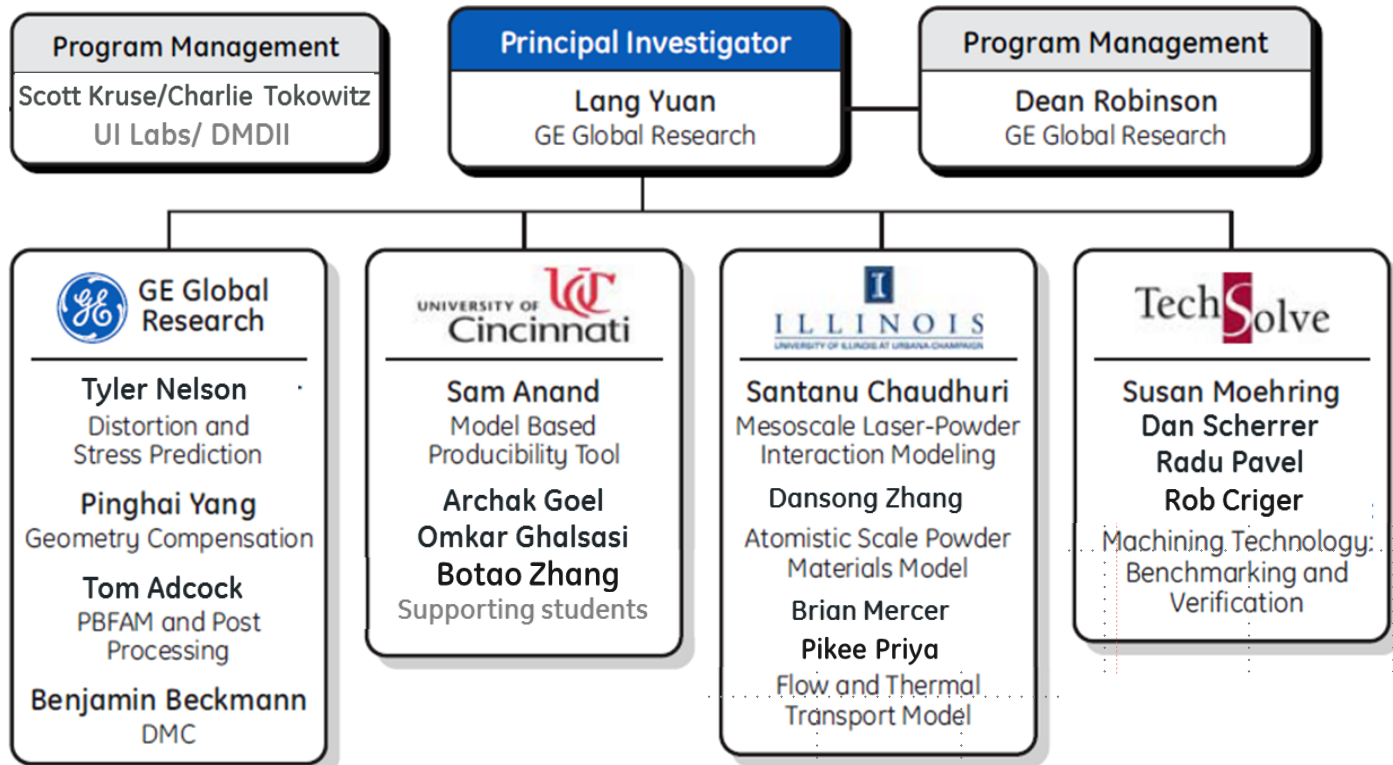
# APS IN-SITU OBSERVATION OF AM PROCESS



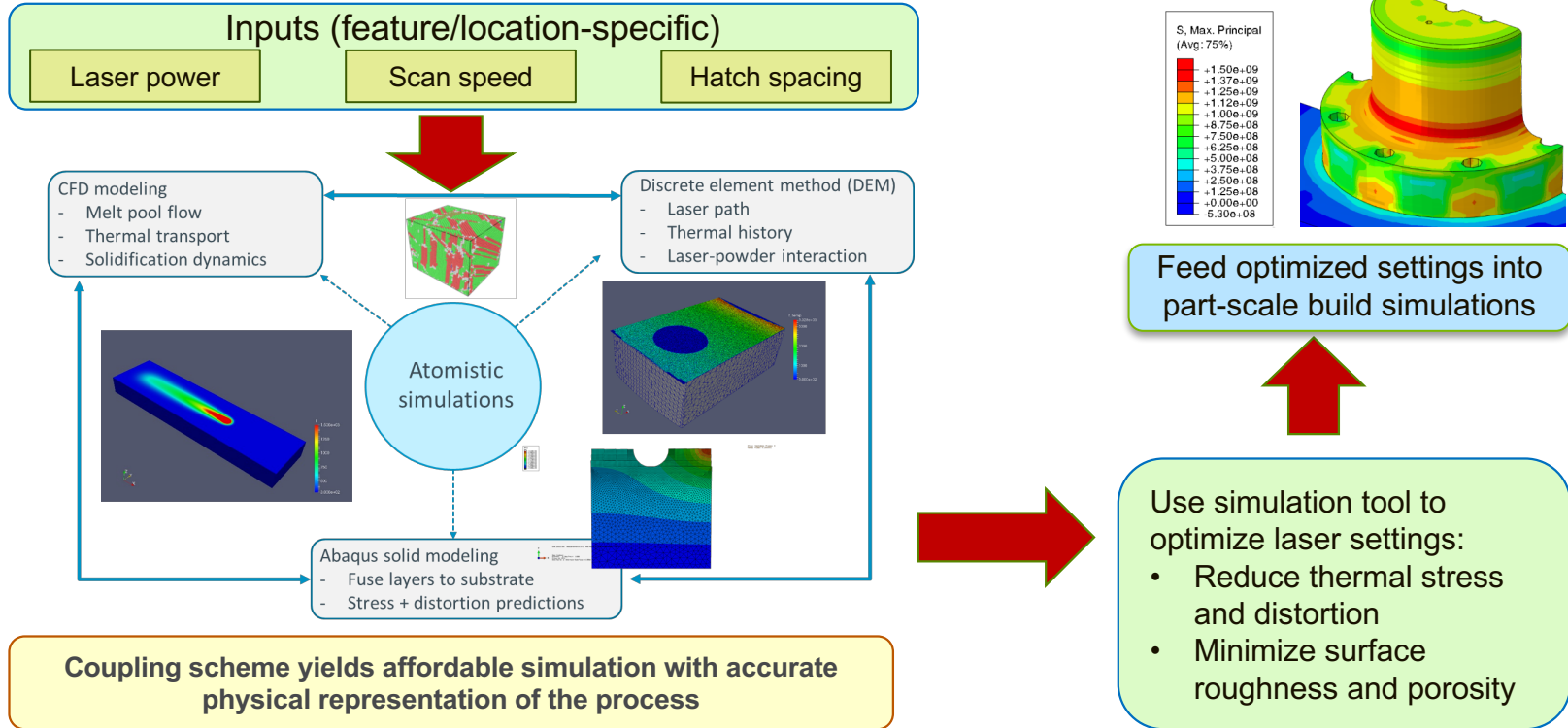
PI – Tao Sun

In-Situ x-ray observation  
Side view | 8ms (45 kHz)

# AM: ACCELERATING QUALIFICATION

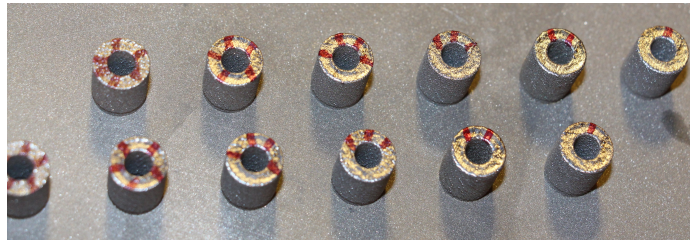


# PROCESS PARAMETER OPTIMIZATION VIA MULTIPHYSICS MESOSCALE SIMULATION



# INFLUENCE OF PROCESS PARAMETERS ON MELT POOL CHARACTERISTICS

| Case | Power (W) | Speed (mm/s) | Pool width (microns) | Pool depth (microns) | Max pool velocity (cm/s) | Max temperature gradient (K/cm) | Max cooling rate (K/s) |
|------|-----------|--------------|----------------------|----------------------|--------------------------|---------------------------------|------------------------|
| 1    | 300       | 900          | 273                  | 55                   | 238                      | 7.10E+05                        | 1.20E+07               |
| 2    | 300       | 600          | 345                  | 60                   | 253                      | 7.30E+05                        | 1.40E+07               |
| 3    | 280       | 1000         | 226                  | 50                   | 214                      | 6.10E+05                        | 1.00E+07               |
| 4    | 250       | 800          | 250                  | 50                   | 210                      | 5.80E+05                        | 9.20E+06               |
| 5    | 200       | 700          | 226                  | 50                   | 190                      | 5.40E+05                        | 7.60E+06               |
| 6    | 200       | 1000         | 178                  | 40                   | 175                      | 4.40E+05                        | 8.00E+06               |



Samples are under characterization on SEM & EBSD

Data above used for microstructure prediction

- Grain Morphology
- Texture

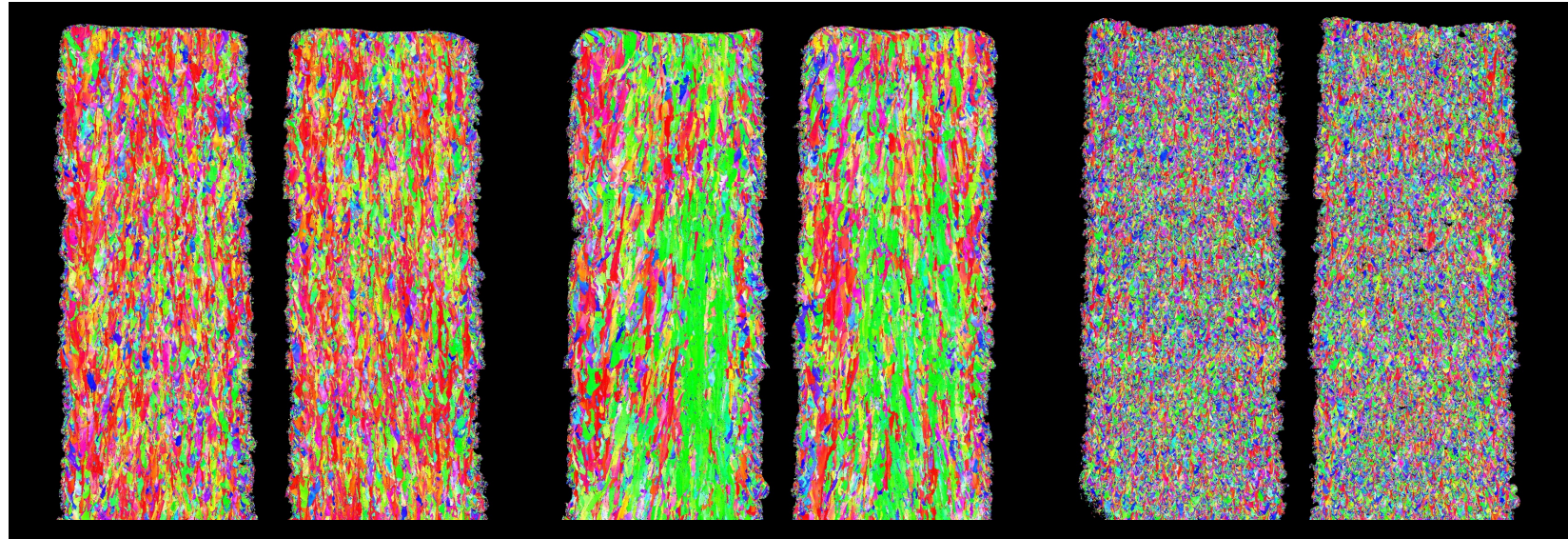


# MICROSTRUCTURE AND PERFORMANCE

1

2

6



1mm

Massive Differences in Microstructure in as-printed parts

# Atmospheric Plasma for Surface Modification and Nanoscale Embedding of Chemistry

Santanu Chaudhuri  
University of Illinois



# Project Team

## Dr. Santanu Chaudhuri (PI)

Professor of Materials Engineering, CME, UIC  
Affiliate, Applied Research Institute (ARI, UIUC)  
Specialist in coatings, corrosion and multiscale modeling

## Dr. David Ruzic

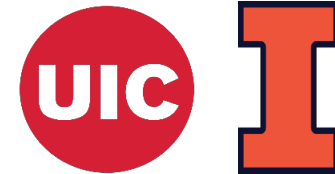
Professor, Nuclear and Plasma Engineering, UIUC  
Specialist in plasma engineering and instrumentation  
Inventor of the ECAP and LAPCAP processes

## Dr. Dan Krogstad

ARI, UIUC  
Coatings, Characterization and Coordination

## Dr. Joseph Osborne

Formerly with Boeing Company (~30 years)  
Specialist in coating performance,  
corrosion protection and coating certification



## Dr. Jessica Krogstad

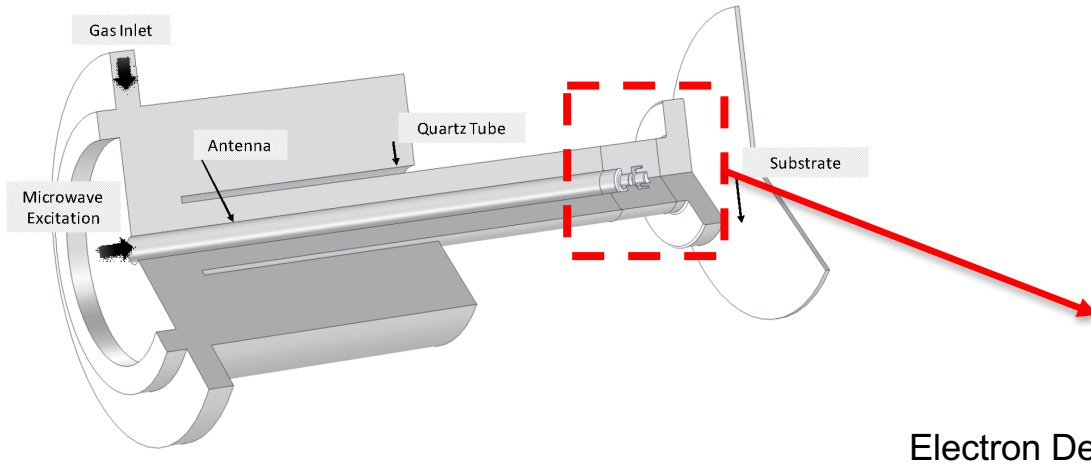
Professor, MatSE, UIUC  
Coating optimization and  
characterization

## Dr. Brian Jurczyk (Starfire Industries)

Small business partner in  
technology assessment,  
transfer, and lifecycle cost

# Atmospheric Pressure Microwave Plasma

- Microwave Excitations: 2.45[GHz]
- Atmospheric Pressure: 760[torr]
- Deposited Power: 200[W]
- Mass flow: 20[lpm]
- Gas: Argon

 $\times 10^{16} [1/m^3]$ 

1

1

# Physics Overview

- Plasma
  - Electron drift-diffusion continuity equation
  - Electron energy equation
  - Electron heat source from microwaves
- Electromagnetics
  - Maxwell equations
  - Plasma-EMW electrical conductivity coupling
- Turbulent flow
  - $k - \epsilon$
- Transient heat transfer
  - Energy equation
  - Heat source form plasma reactions

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{V}) = 0$$

$$\rho \frac{\partial \mathbf{V}}{\partial t} + \rho (\mathbf{V} \cdot \nabla) \mathbf{V} = \nabla \cdot (-p \hat{\mathbf{I}} + \hat{\boldsymbol{\tau}}) + \mathbf{F}$$

Lorentz Force:  $\mathbf{F} = \frac{1}{2} \mathcal{R}e\{\sigma \mathbf{E} \times \mathbf{B}^*\}$

$$\rho C_p \frac{\partial T}{\partial t} + \rho C_p \mathbf{V} \cdot \nabla T = \nabla \cdot (\lambda \nabla T) + Q$$

$$N_i \frac{\partial}{\partial t} (\ln N_i) + \nabla \cdot [-D_i N_i \nabla (\ln N_i) + N_i \mathbf{V}] = S_i$$

**Reactive Flow**

Faraday's law:  $\nabla \times \mathbf{E} = -j\omega \mu_0 \mathbf{H}$

Ampère's law:  $\nabla \times \mathbf{H} = -j\omega \epsilon_0 \epsilon_p \mathbf{E}$

Electric Permittivity:  $\epsilon_{pl} = 1 - \frac{\omega_p^2}{\omega(\omega - j\nu m)}$

Plasma Frequency:  $\omega_p = \left(\frac{n_e e^2}{\epsilon_0 m_e}\right)^{1/2}$

Electric Conductivity:  $\sigma = \frac{n_e e^2}{m_e \nu m + j\omega}$

Resistive Heating:  $Q_h = \frac{1}{2} \mathcal{R}e\{\sigma \mathbf{E} \cdot \mathbf{E}^*\}$

**Microwave**

Boltzmann:  $\frac{\partial}{\partial t} (Wf - D \frac{\partial f}{\partial \epsilon}) = S$

$$W = -\gamma \epsilon^2 \sigma_\epsilon - 3a \left(\frac{n_e}{N_n}\right) A_1$$

$$D = \frac{\gamma}{3} \left(\frac{E}{N_n}\right)^2 \left(\frac{\epsilon}{\sigma_m}\right) + \frac{\gamma k_b T}{q} \epsilon^2 \sigma_\epsilon + 2a \left(\frac{n_e}{N_n}\right) (A_2 + \epsilon^2 A_3)$$

$$\frac{\partial n_e}{\partial t} + \nabla \cdot [-n_e (\mu_e \mathbf{E}) - \mathbf{D}_e \nabla n_e] = R_e$$

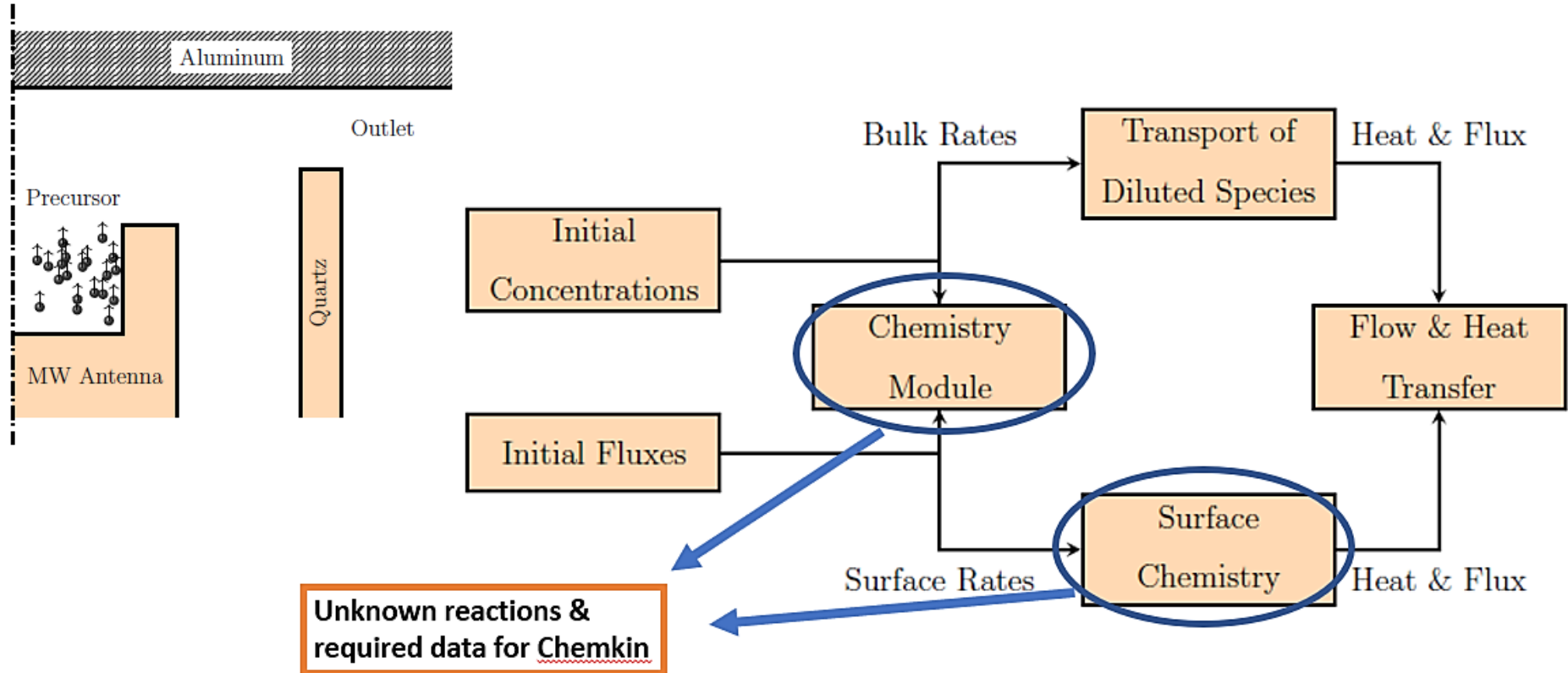
Electron Source:  $R_e = \sum x_j k_j N_n n_e$

$$\frac{\partial n_e}{\partial t} + \nabla \cdot [-n_e (\mu_e \mathbf{E}) - \mathbf{D}_e \nabla n_e] + \mathbf{E} \cdot \Gamma_e = R_e$$

Inelastic Collision energy loss:  $R_e = \sum x_j k_j N_n n_e \Delta \epsilon_j$

**Plasma**

# Multiscale Chemistry



# Surface cleaning: Nature of the Reactions

The key mechanisms occurring are

- Evaporation
- Thermal Degradation
- Oxidative Volatility
- Polymerization
- Deposit Formation
- Oxidation
  - Initiation
  - Propagation
  - Branching
  - Termination

lubricating oils

- C<sub>n</sub>H<sub>2n-2</sub> to C<sub>n</sub>H<sub>2n-20</sub>

Pennsylvania oil is generally agreed to be free from unsaturation

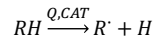
- Pennsylvania oil has been the most studied because of its relatively less complex composition.
- C<sub>n</sub>H<sub>2n-8</sub> to C<sub>n</sub>H<sub>2n-20</sub>
- Naphthenes (cycloalkanes C<sub>n</sub>H<sub>2(n+1-r)</sub>)
  - n: carbon
  - r: rings
- Paraffin or iso-paraffin side chains
- n<sub>C</sub> = 0.358 mol. wt. + 7.7 - 0.3 mol. vol.

Initiation: ROOH → RO<sub>2</sub> + RO + OH

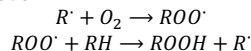
Propagation: RO<sub>2</sub> + RH → RO<sub>2</sub>H + R      R + O<sub>2</sub> → RO<sub>2</sub>

## Primary Free Radical Oxidation

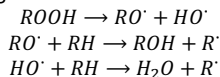
Initiation:



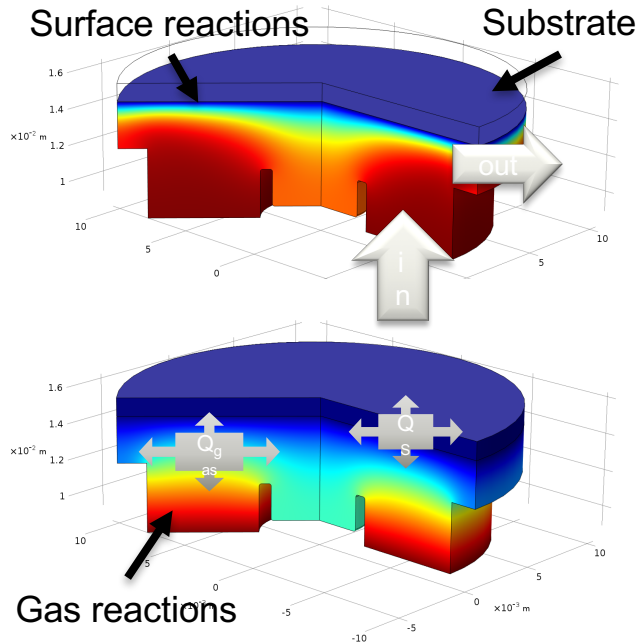
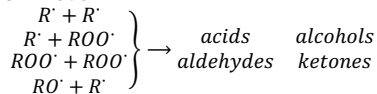
Propagation:



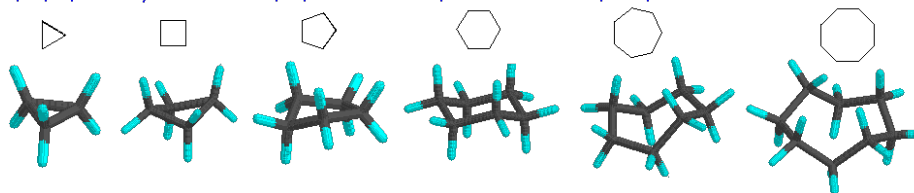
Branching



Termination



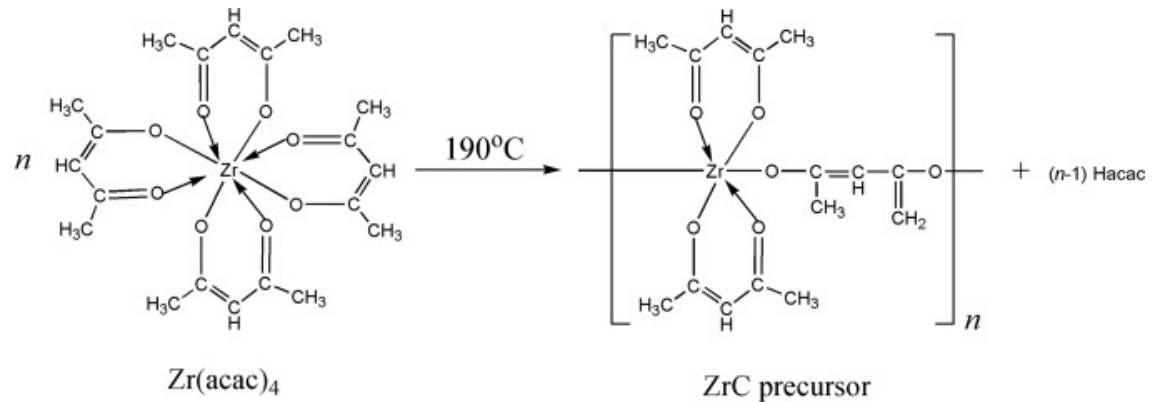
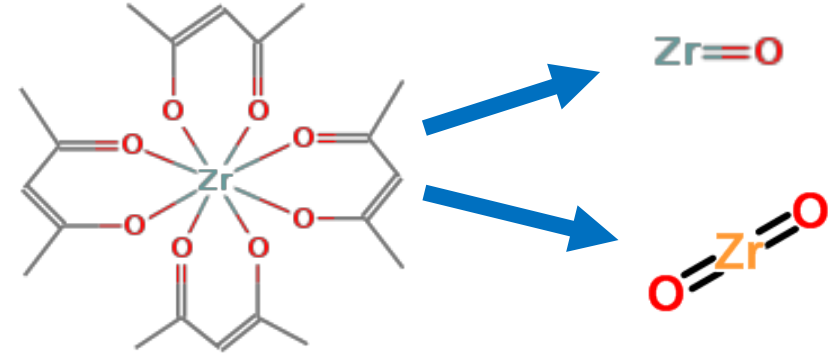
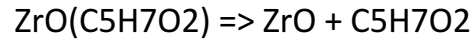
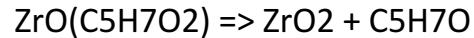
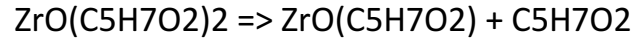
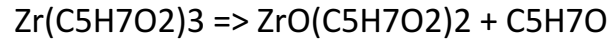
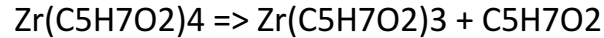
cyclopropane    cyclobutane    cyclopentane    cyclohexane    cycloheptane    cyclooctane



# Reaction Engineering

- In the APP process, Zirconium Acetylacetonate ( $Zr(C_5H_7O_2)_4$ ) first decomposes into a gas phase. The reaction products along with water vapor ( $H_2O$ ) adsorb and react on an Aluminum (Al) substrate to form Zirconium Oxide ( $ZrO_2$ ) layers.

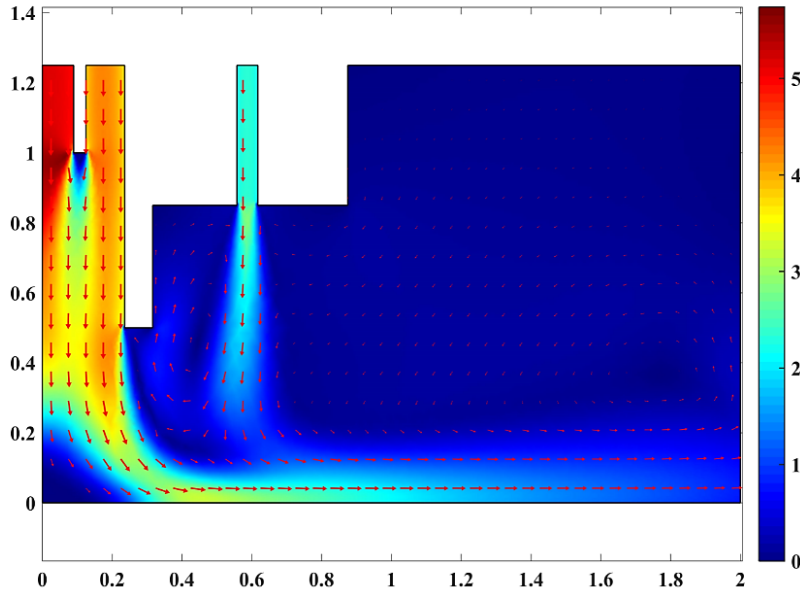
## 1. Gas-phase decomposition reactions



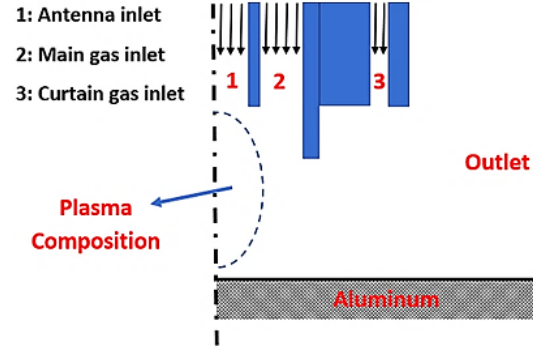


# ZrO<sub>2</sub>: Optimizing Flow Rate and Growth

Curtain gas flow rate = 20 L/min, Velocity magnitude (m/s)

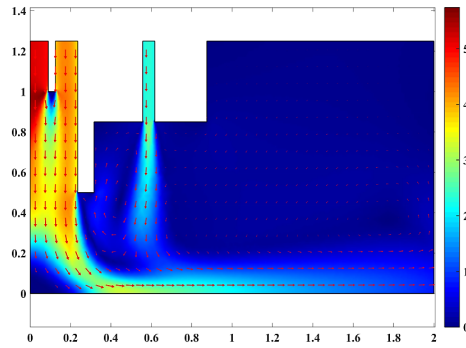


Precursor:  
 $Zr(AcAc)_4$   
 $U_1 = 5$  l/min  
 $U_2 = 20$  l/min

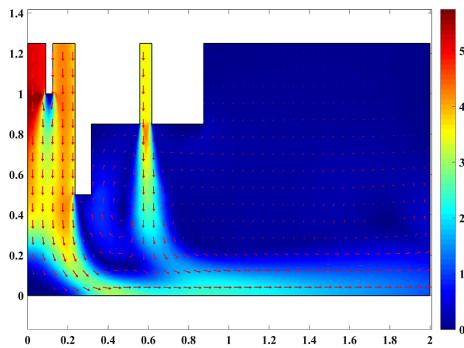


- ❑ Considering the turbulence effects in the models, the flow pattern is smoother at lower gas-curtain flow rate (20 – 40 lpm) almost without eddies.
- ❑ The growth rate of zirconia is higher at lower flow rates due to almost laminar flow pattern which increases the residence time for ZrO and Zr to participate in surface reactions.

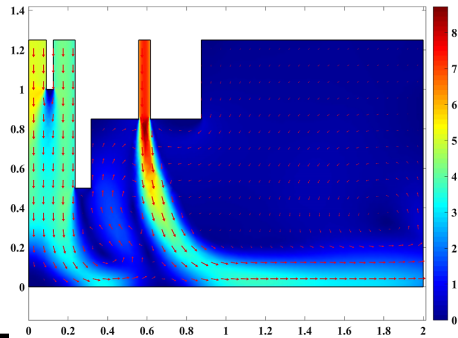
Curtain gas flow rate = 20 L/min, Velocity magnitude (m/s)



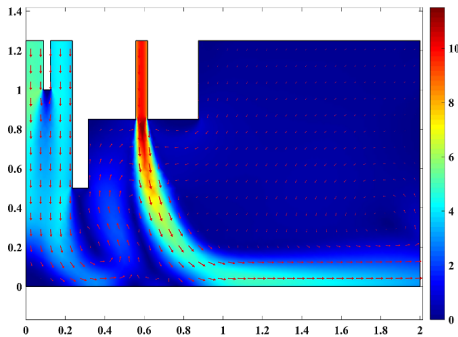
Curtain gas flow rate = 30 L/min, Velocity magnitude (m/s)



Curtain gas flow rate = 60 L/min, Velocity magnitude (m/s)



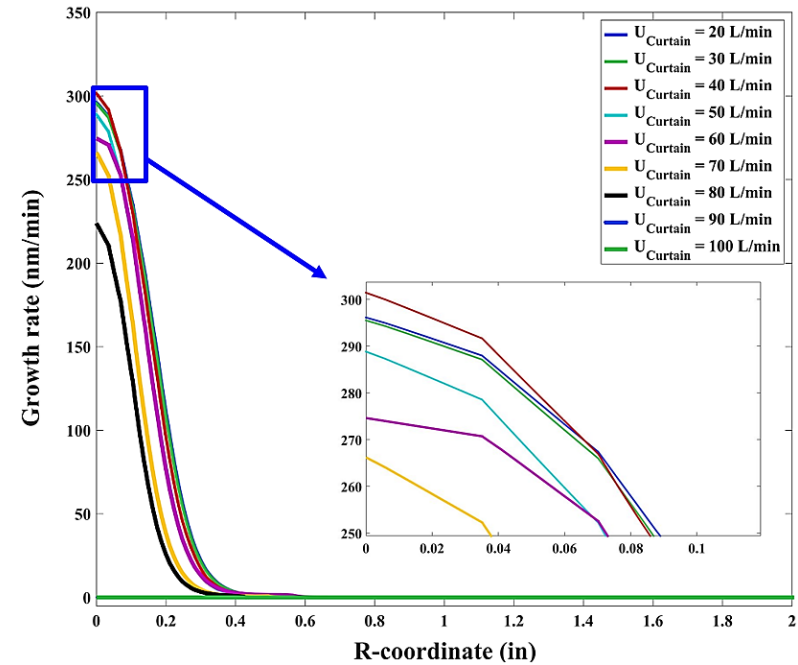
Curtain gas flow rate = 80 L/min, Velocity magnitude (m/s)



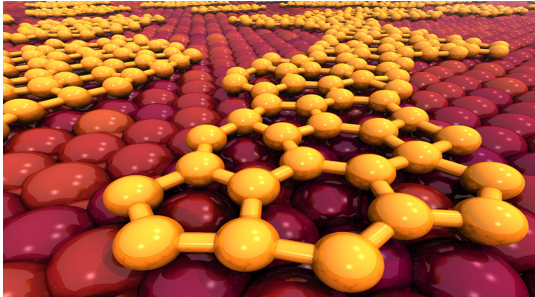
Precursor:  $Zr(AcAc)_4$

$U_1 = 5$  l/min

$U_2 = 20$  l/min



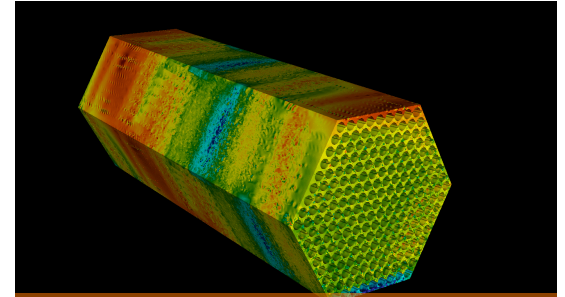
# SUMMARY: PREDICTIVE CONTROL OVER NANOSCALE DOMAINS IN 3D



**MULTISCALE LINKS  
TO MATERIALS  
PHYSICS AND  
CHEMISTRY**



**MESOSCALE  
MODELS, AI/ML  
AND PROCESS  
MODELING**



**MATERIALS, DEVICE,  
AND FABRICATION TO  
PERFORMANCE,  
RELIABILITY AND  
AGING**

PREDICT

VALIDATE

LEARN AND DECIDE

BUILD