

Kiran Vaddi

Brenden
Pelkie

Huat Thart-
Chiang

Karen Li

Maria
Politi

W
UNIVERSITY *of* WASHINGTON

Integrating ML and AI with 'Open' Laboratory Automation

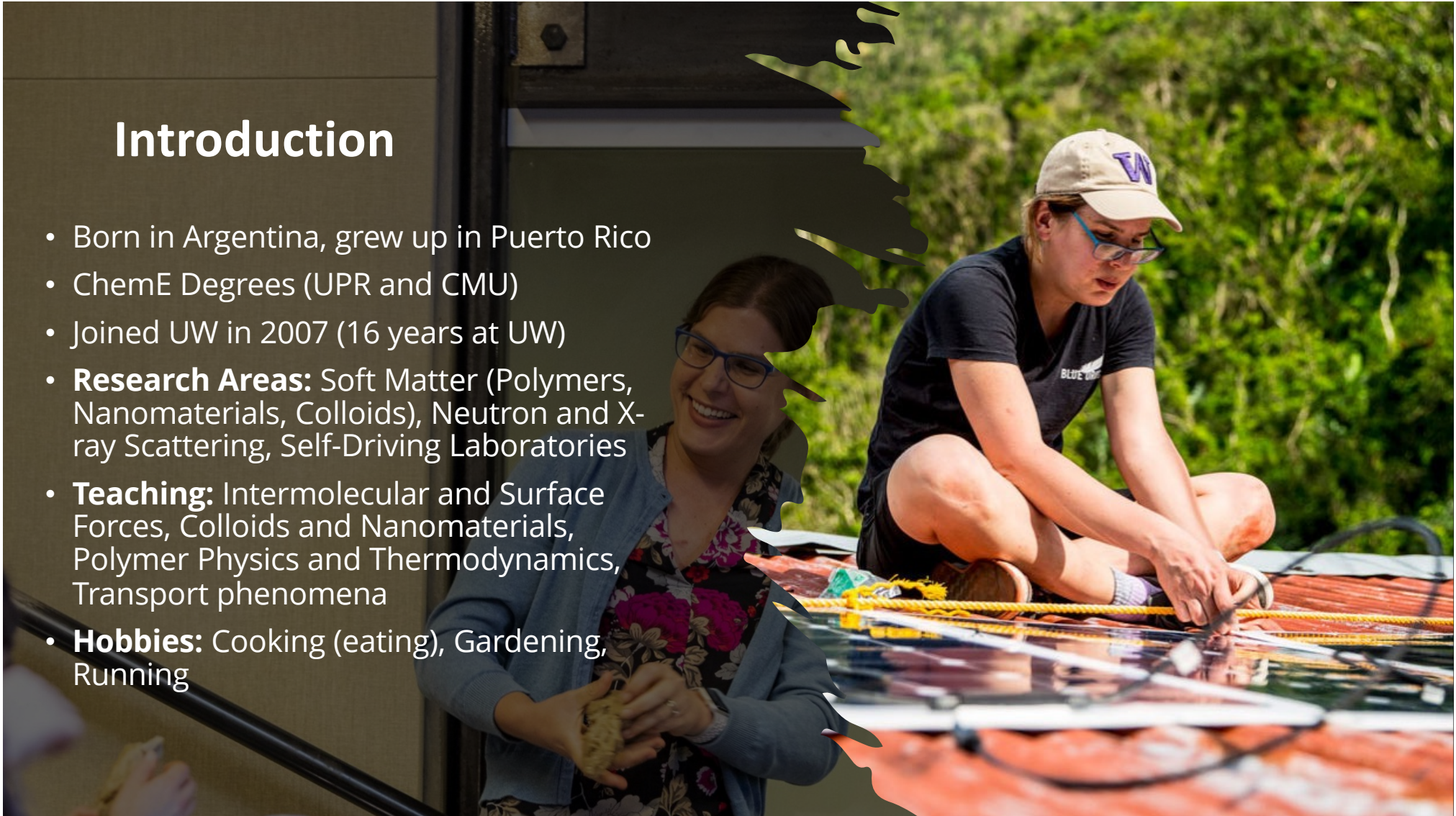


Prof. Lilo D. Pozzo
Chemical Engineering
University of Washington



Introduction

- Born in Argentina, grew up in Puerto Rico
- ChemE Degrees (UPR and CMU)
- Joined UW in 2007 (16 years at UW)
- **Research Areas:** Soft Matter (Polymers, Nanomaterials, Colloids), Neutron and X-ray Scattering, Self-Driving Laboratories
- **Teaching:** Intermolecular and Surface Forces, Colloids and Nanomaterials, Polymer Physics and Thermodynamics, Transport phenomena
- **Hobbies:** Cooking (eating), Gardening, Running

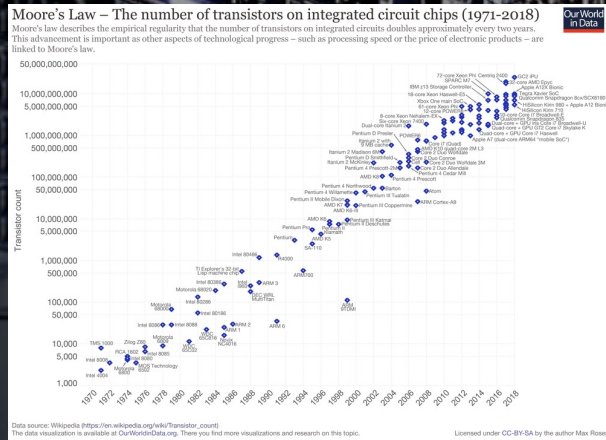




Modern-Day Materials Challenges

- Clean energy
 - Energy conversion
 - Energy storage
- Clean water
- Disease management
- Sustainability
 - Food
 - Polymers
 - Infrastructure

Can we achieve Moore's law for Materials Design?

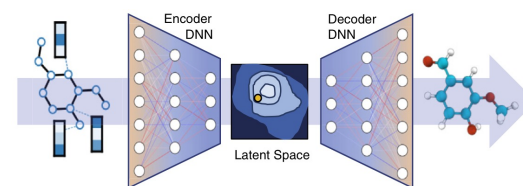
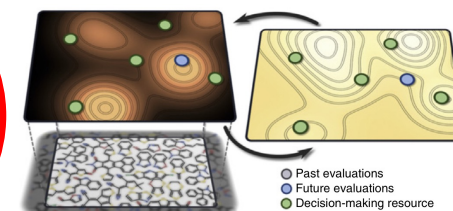
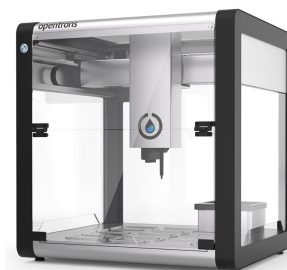
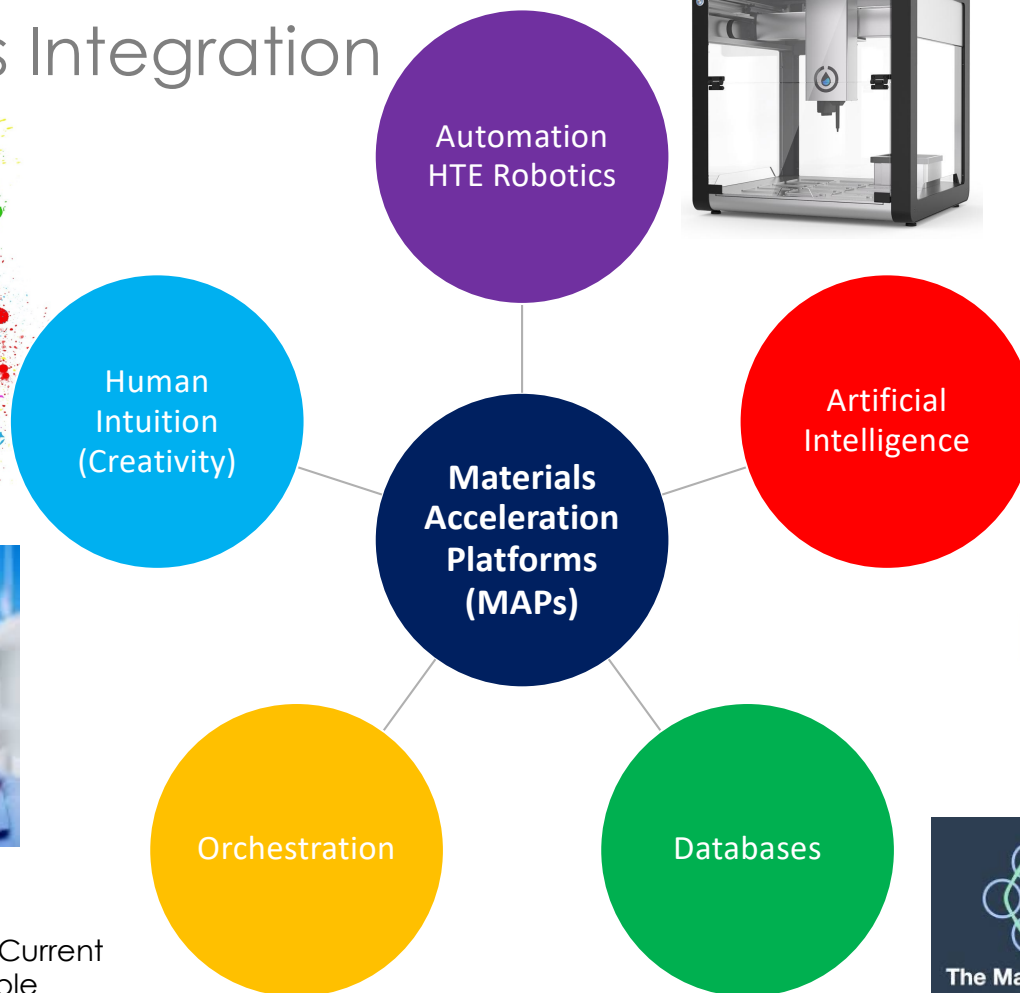


It takes 10-20 years to fully develop new materials

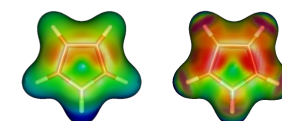
Way too slow...



MAP: Systems Integration



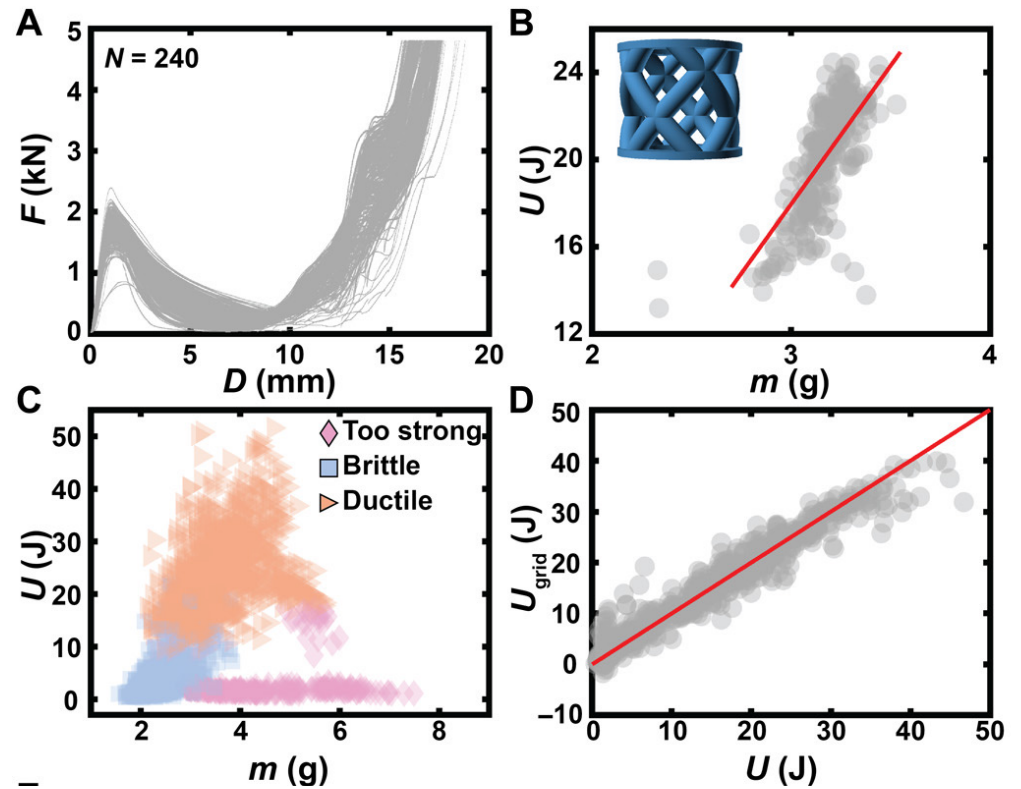
Martha M. Flores-Leonar, et.al. Current Opinion in Green and Sustainable Chemistry, 100370, (2020)



Recent Examples and Implementations

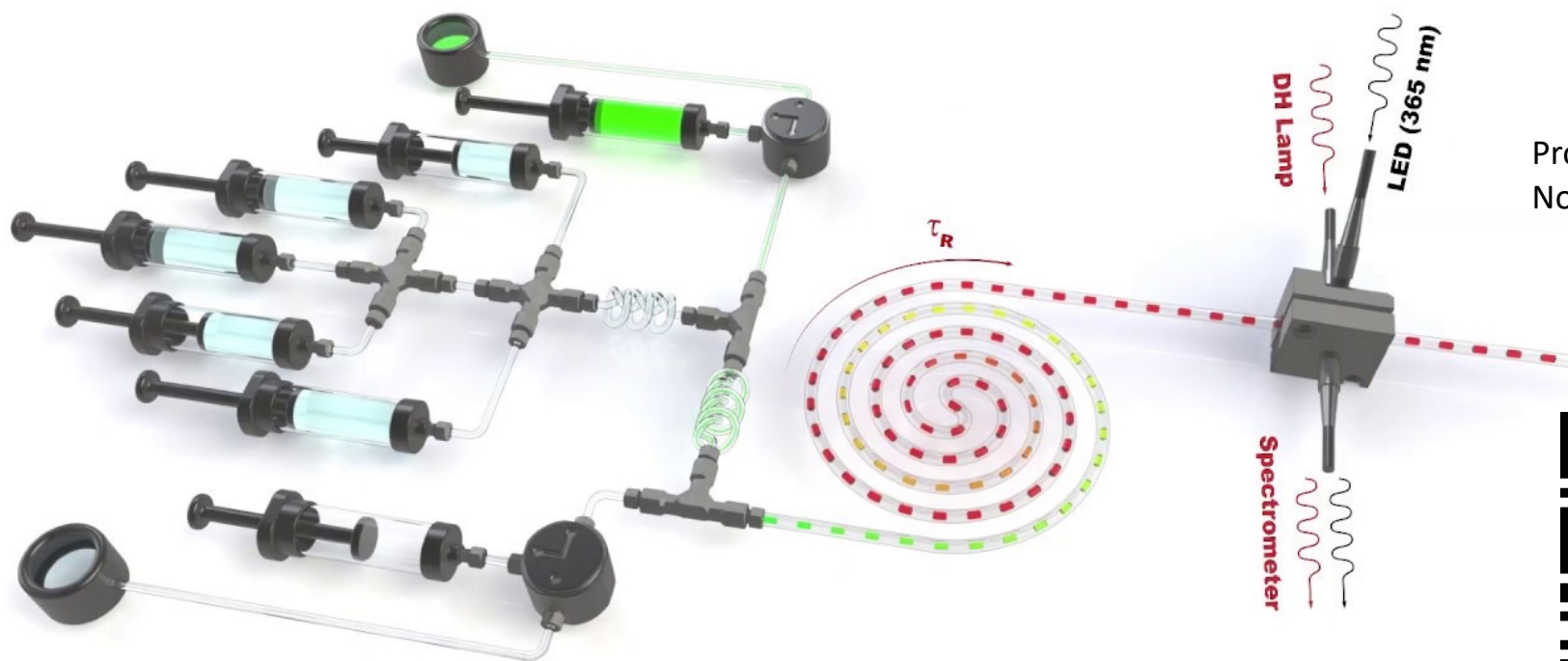


Prof. Keith Brown, Boston University



Recent Examples and Implementations

Experimental Setup

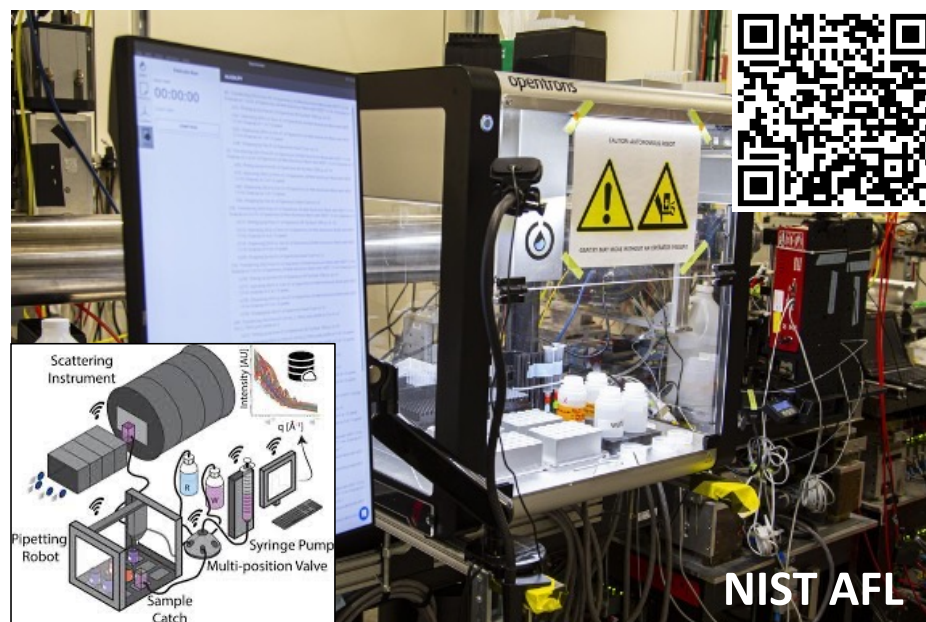
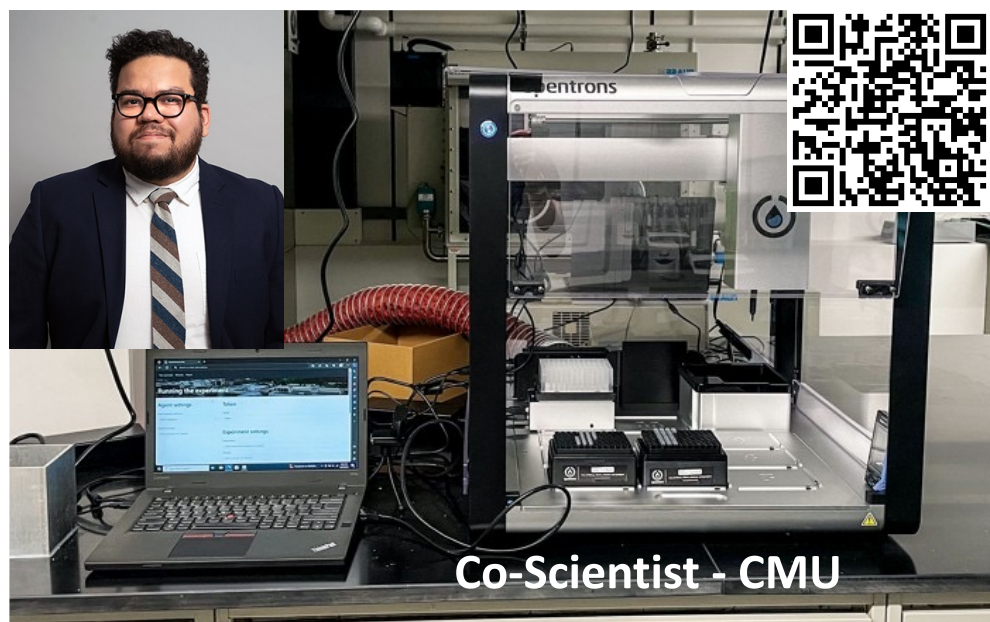


Prof. Milad Abolhasani
North Carolina State U.



OT2: A Success Story in Accessible SDL Platforms

- Low Cost Facilitates Adoption
- Open & Accessible (Python API)
- Adopted Standards (SLAS)
- Limited Capability (Move Liquids)
- Single Tool (Pipettes)
- Increasing Costs



AI-Driven High-Throughput Materials Research

AI Orchestration

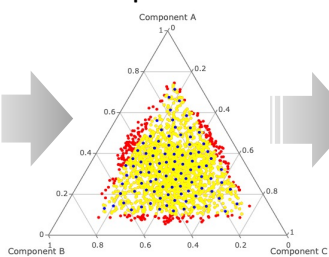
**AI-Driven
Materials
Exploration**

Iterative Exploration Campaigns

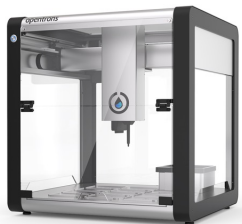
Prior Base
Design Knowledge



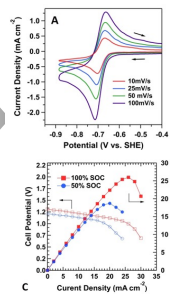
Design
Of Experiments



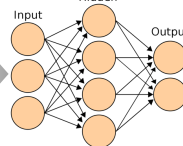
HT
Sample Formulation



HT
Characterization

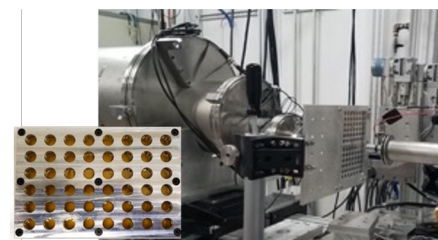
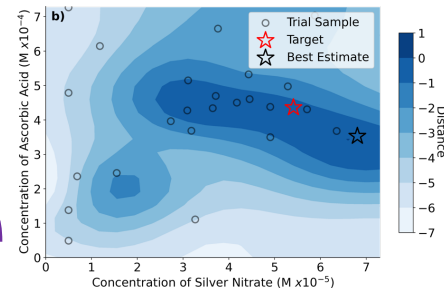


Dimensionality
Reduction
Feature Extraction
Machine Learning



Data / Metadata Stewardship

**Optimized New
Materials**



HT X-ray Characterization



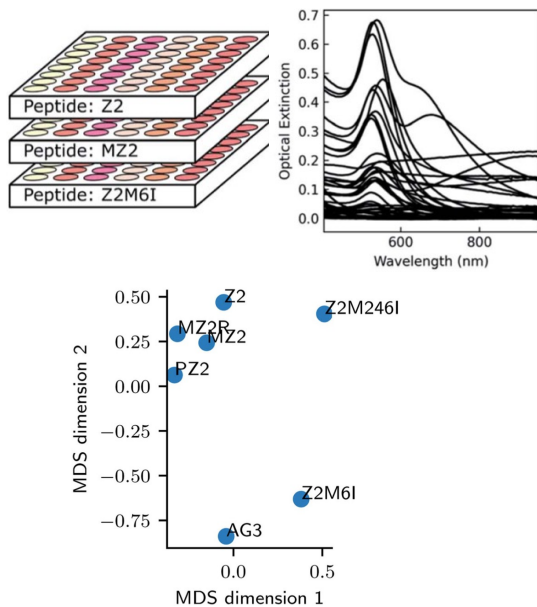
@lilopozzo



Pozzo-Research-Group

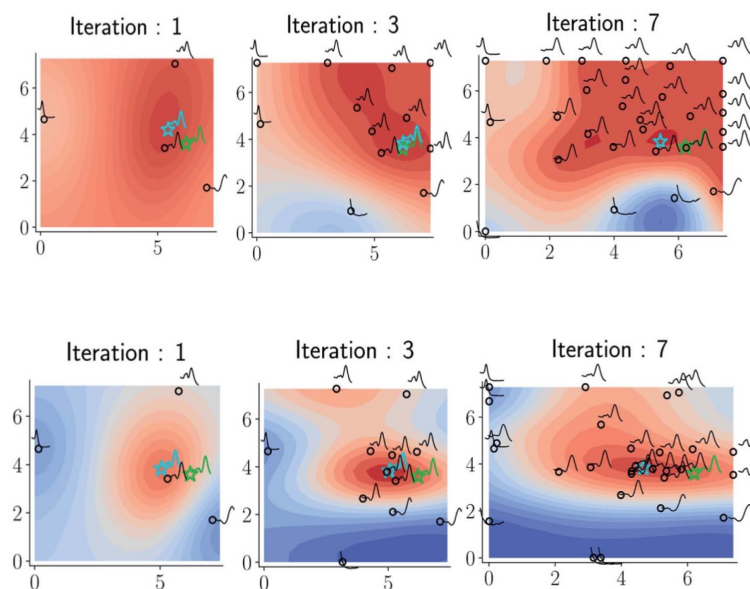
Examples of Materials Research Objectives

Multi-dimensional Correlations



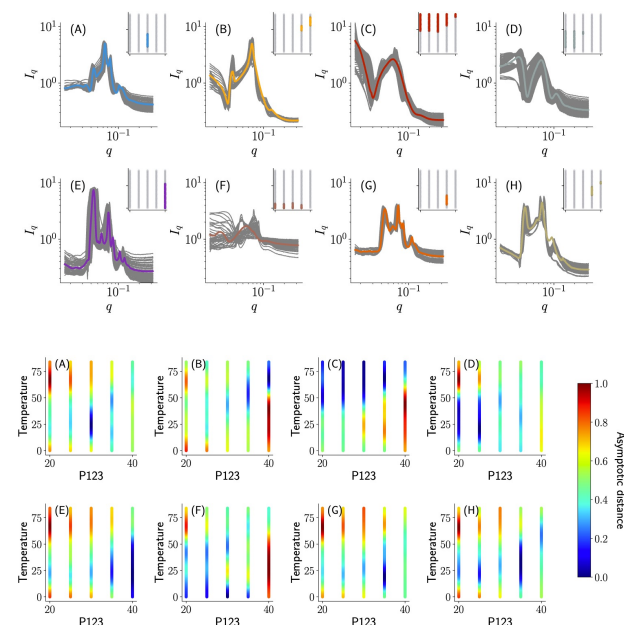
Lachowski, Kacper J., et al. "Multivariate analysis of peptide-driven nucleation and growth of Au nanoparticles." *Digital Discovery* (2022).

Retrosynthesis



Vaddi, Kiran, Huat Thart Chiang, and Lilo D. Pozzo. "Autonomous retrosynthesis of gold nanoparticles via spectral shape matching." *Digital Discovery* (2022).

Phase Mapping



K. Vaddi, K. Li, L. Pozzo, "Metric geometry tools for automatic structure phase map generation", *Digital Discovery*, 2, 1471-1483, (2023)

Open Questions

- How to know if target is 'synthesizable'?
- Is the design space large enough? Too large?
- How to integrate established physical knowledge in AI?
- Synergistically use experiments and simulations?
- Can we discover new physical concepts?

Broadening Access

Democratizing MAPs and SDLs

How will the lab of the future look like?

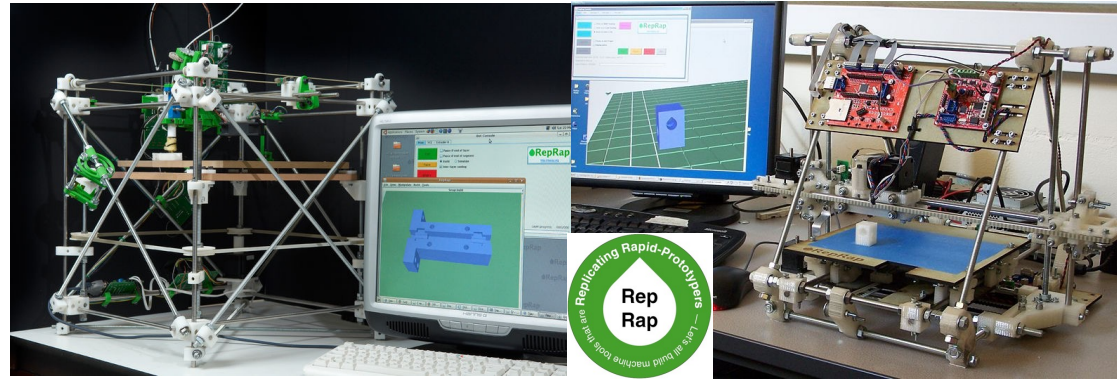
Prof. Andy Cooper



1. Who can access these tools?
2. Who is trained to use them? ... Who develops them?
3. Will they be broadly adopted?
4. Will SDLs revolutionize every field?
5. Who will they serve?

Accelerating Development and Access: Open Hardware

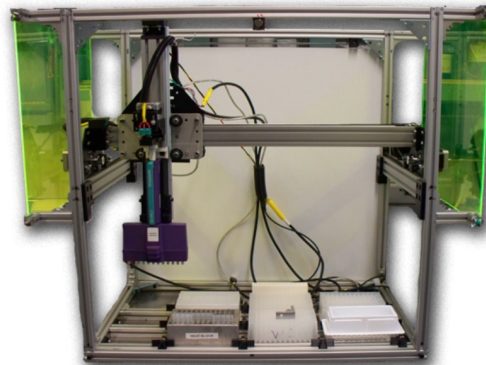
Open-Source RepRap Project (from 2005)



Commercial



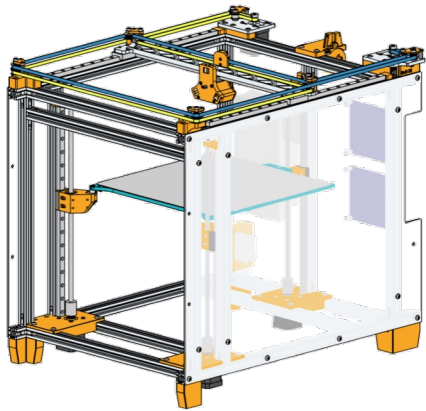
OT-One (~2014)



OT2 (~2018)



Jubilee: an open-source, multi-tool motion platform



- Automatic tool change capabilities
- Stemming from the Maker Movement
- Open-hardware design- reconfigurable
- Affordable : ~\$2000 USD (motion platform)

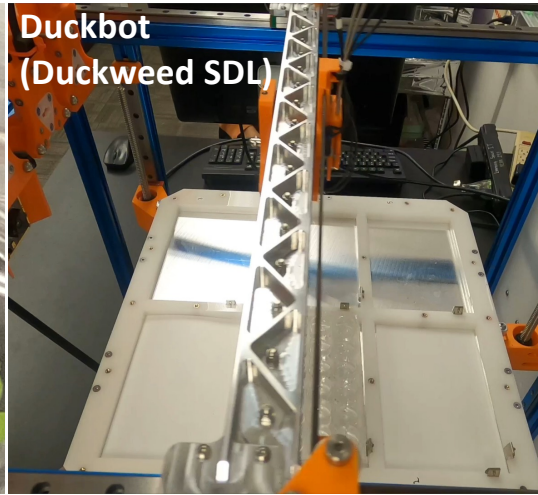
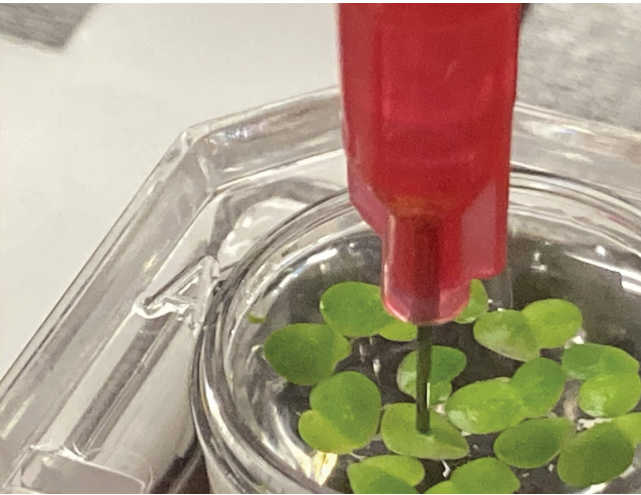
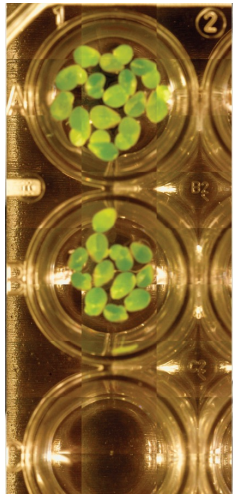


Prof. Nadya Peek
Machine Agency



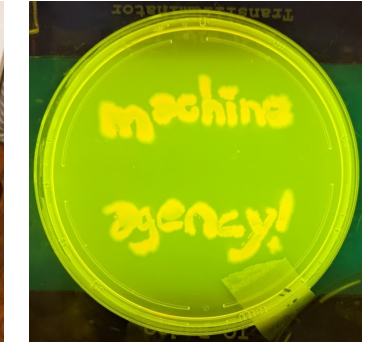
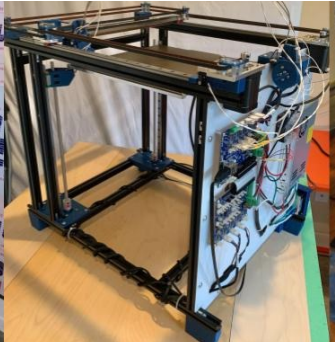
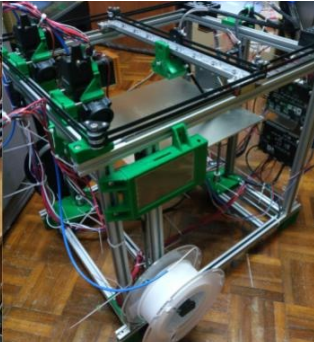
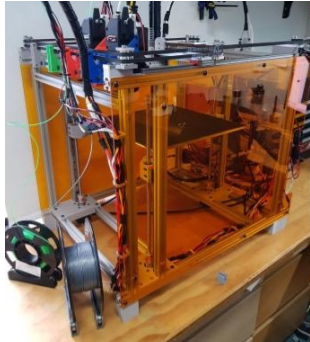
- Accessible
- Allows creation of automation ecosystems
- Enables broader tool development/libraries

Opportunities for Open-Hardware in Lab Automation



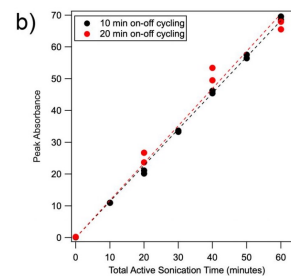
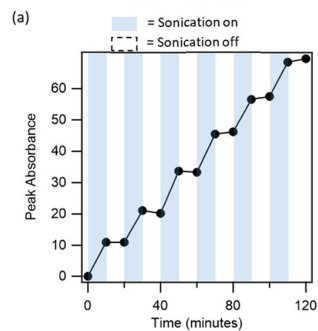
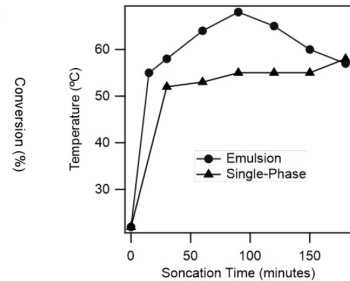
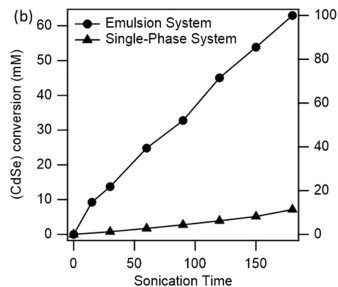
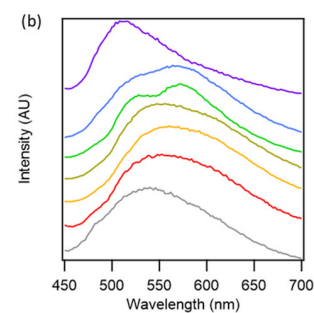
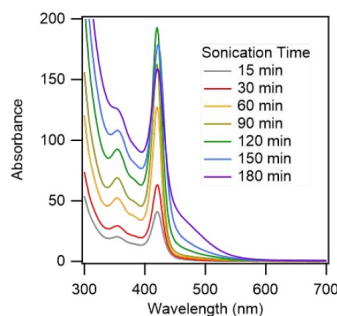
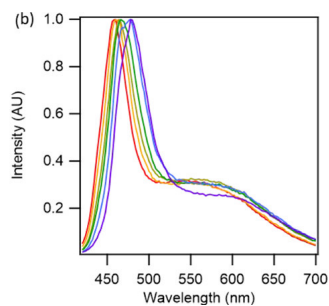
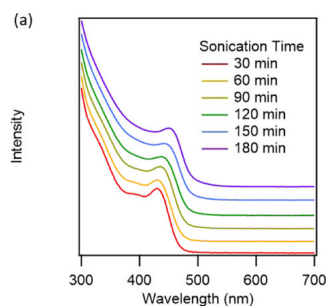
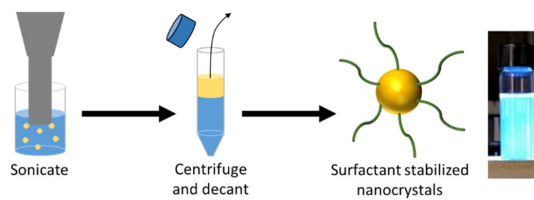
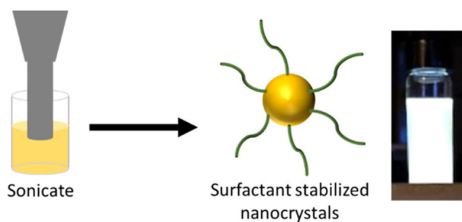
Duckbot
(Duckweed SDL)

Jubilee for Science



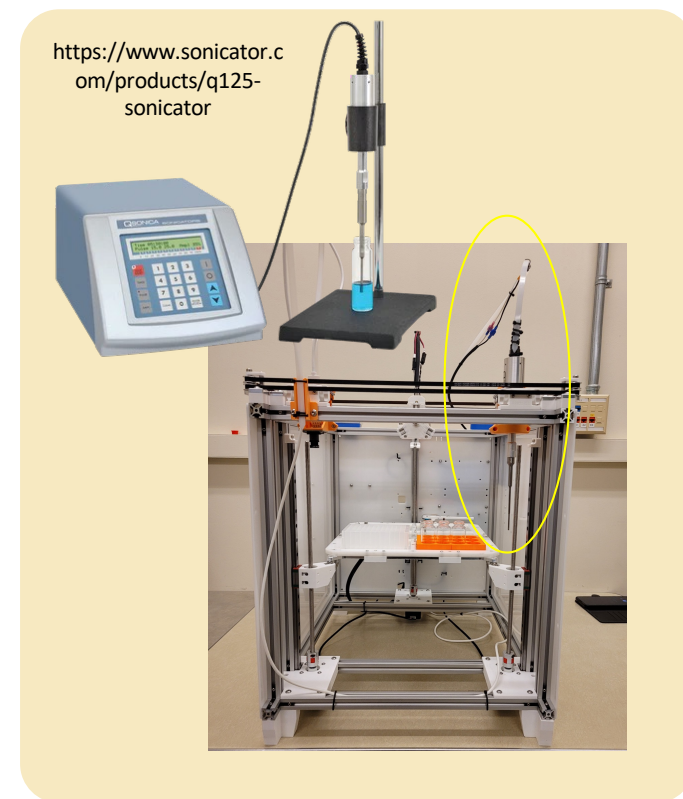
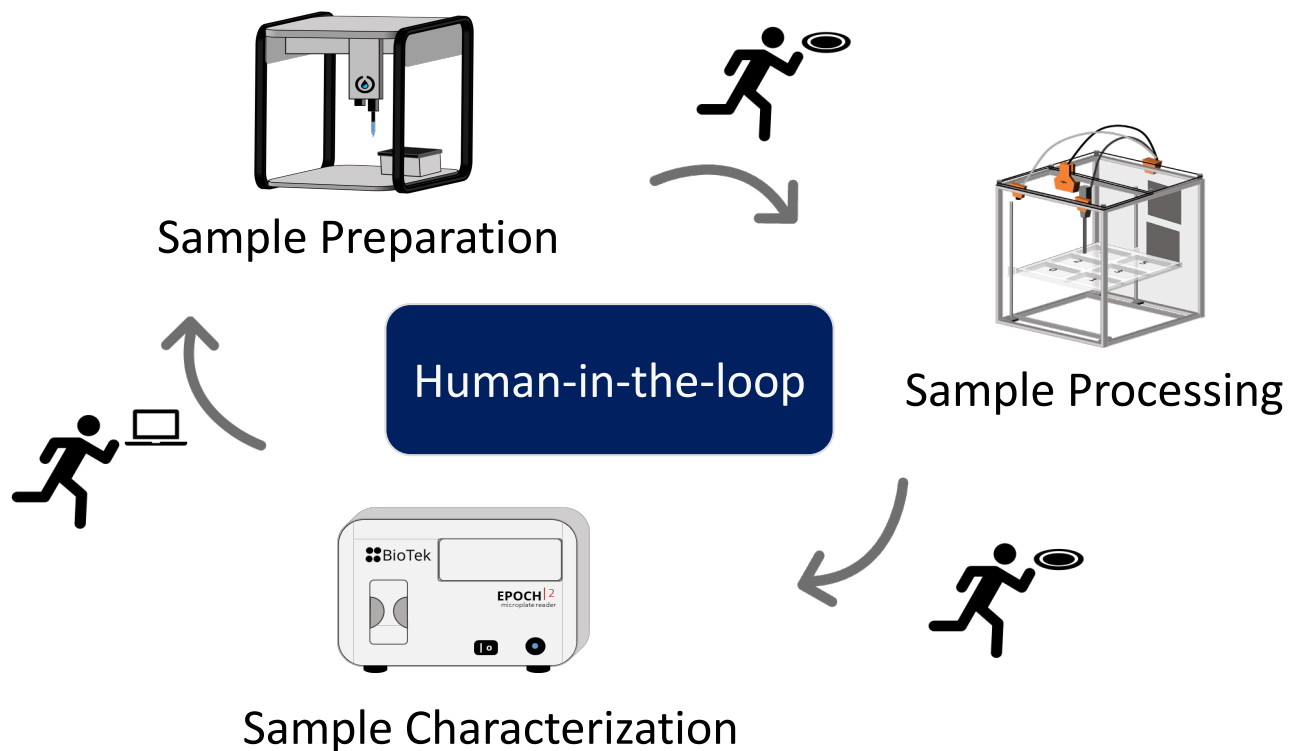
The Duckbot: A system for automated imaging and manipulation of duckweed B Subbaraman, O de Lange, S Ferguson, N Peek. Plos one 19 (1), e0296717 (2024)

Sonochemical Synthesis of CdSe Qdots and Magic-Sized Clusters



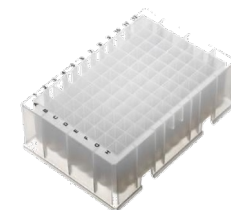
Kastilani, R.; Bishop, B. P.; Holmberg, V. C.; Pozzo, L. D., *Langmuir*, 2019, 35, 16583-16592.

Jubilee for science – sonochemical CdSe QDs



“A high-throughput workflow for the synthesis of CdSe nanocrystals using a sonochemical materials acceleration platform” M. Politi², F. Baum, K. Vaddi, E. Antonio¹, J. Vasquez, B. P Bishop, N. Peek, V. C Holmberg, LD Pozzo, Digital Discovery, 2, 1042-1057 (2023)

Preliminary Campaign for Colloidal CdSe



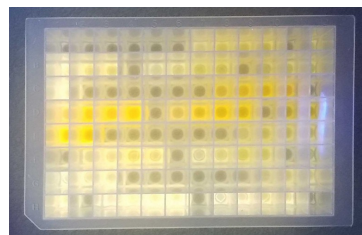
4 Different Components

5 Concentrations

625 samples (triplicate)

0.5 mL total volume/sample

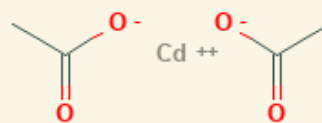
5 minutes of (active) Sonication



In Octadecene

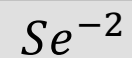
Metal Precursors

Cadmium Acetate



(Stock solution in TOP)

Selenium

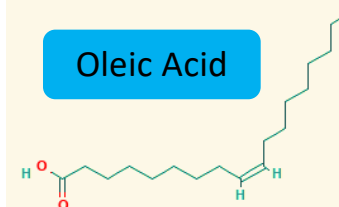


Anionic
Selenide

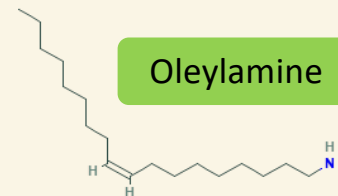
Ligands

(Pure Stocks)

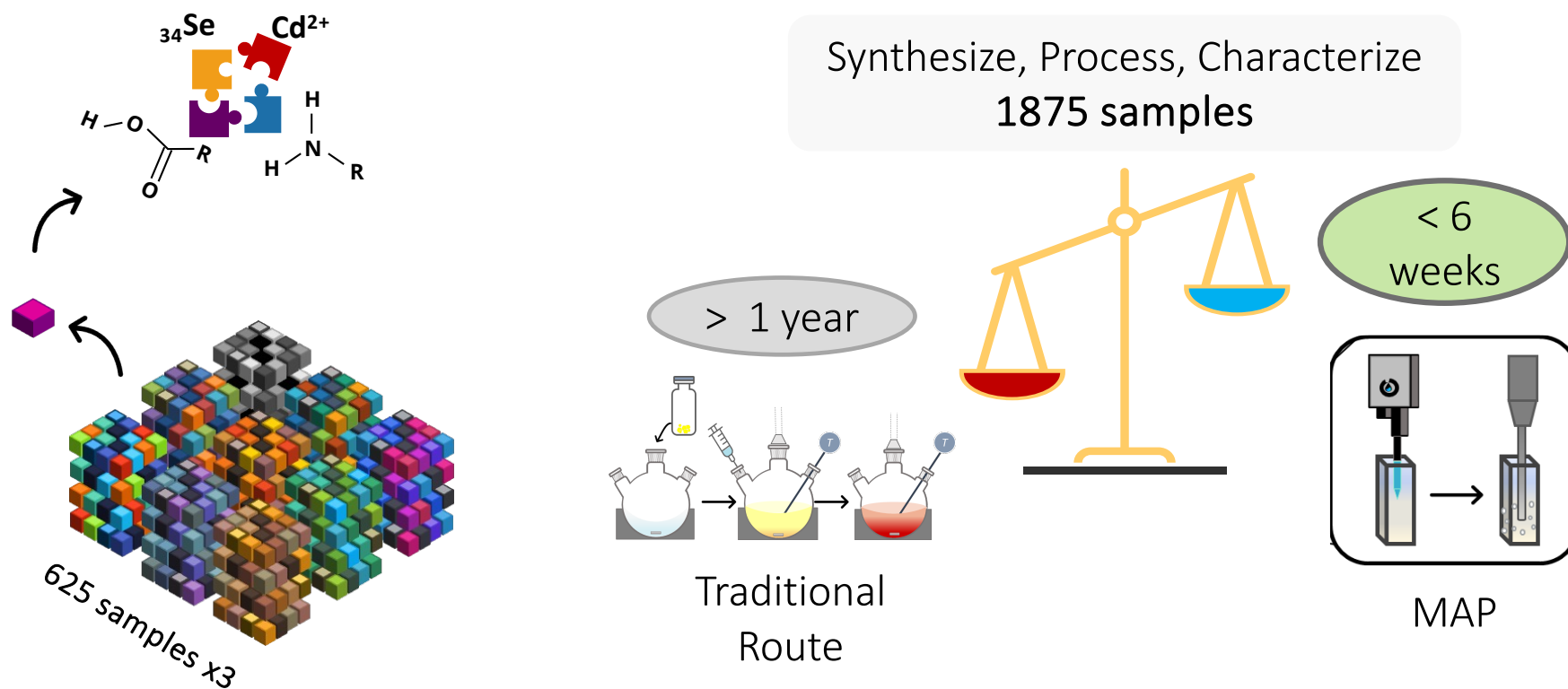
Oleic Acid



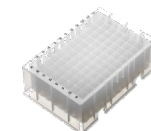
Oleylamine



Experimental space explored in less than 6 weeks

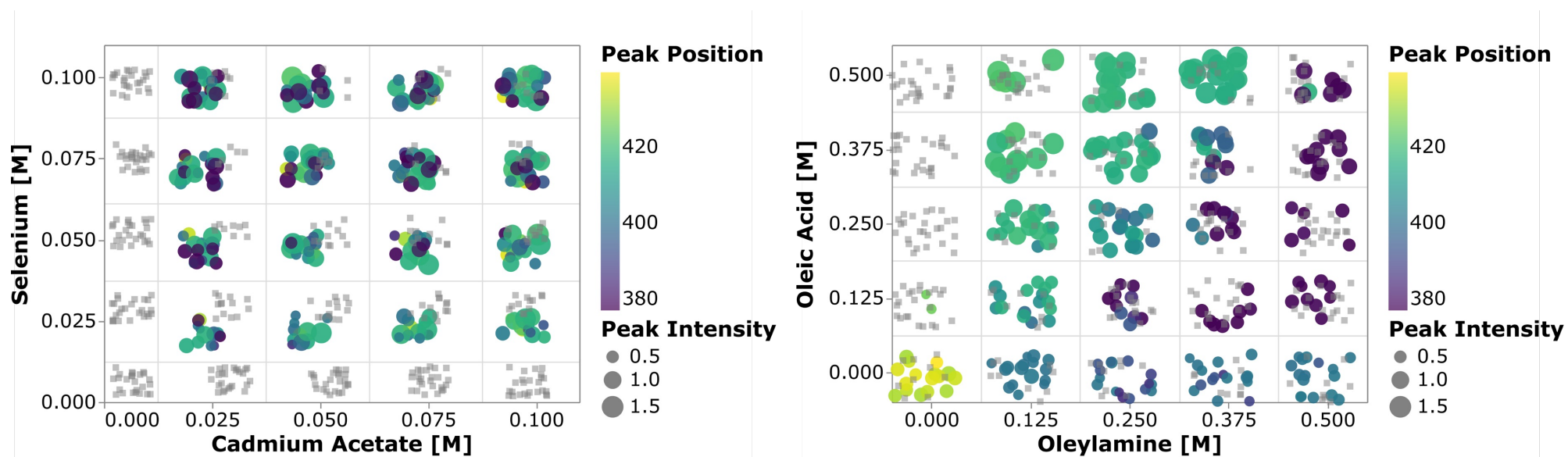


“A high-throughput workflow for the synthesis of CdSe nanocrystals using a sonochemical materials acceleration platform” M. Politi², F. Baum, K. Vaddi, E. Antonio¹, J. Vasquez, B. P Bishop, N. Peek, V. C Holmberg, LD Pozzo, Digital Discovery, 2, 1042-1057 (2023)

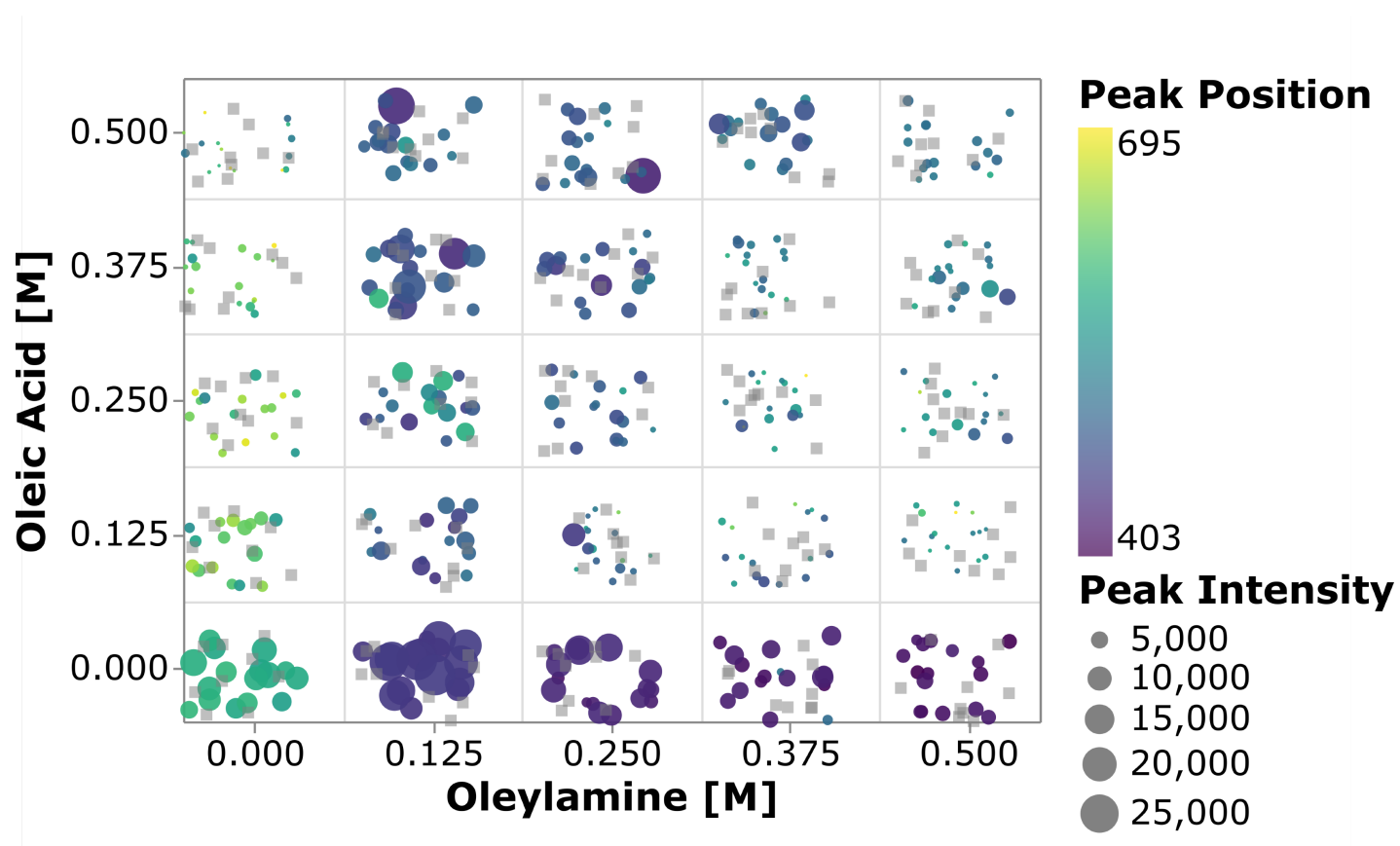
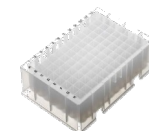


UV-Vis Extinction: Large Data Representation

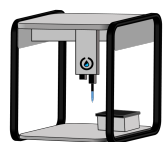
Each box: experimental condition for the ligands
Circles are randomly shifted to avoid overlapping



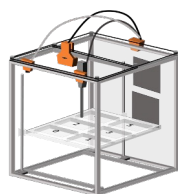
PL Spectroscopy Large Data Representation



Jubilee as a fully integrated self-driving lab



Sample Preparation



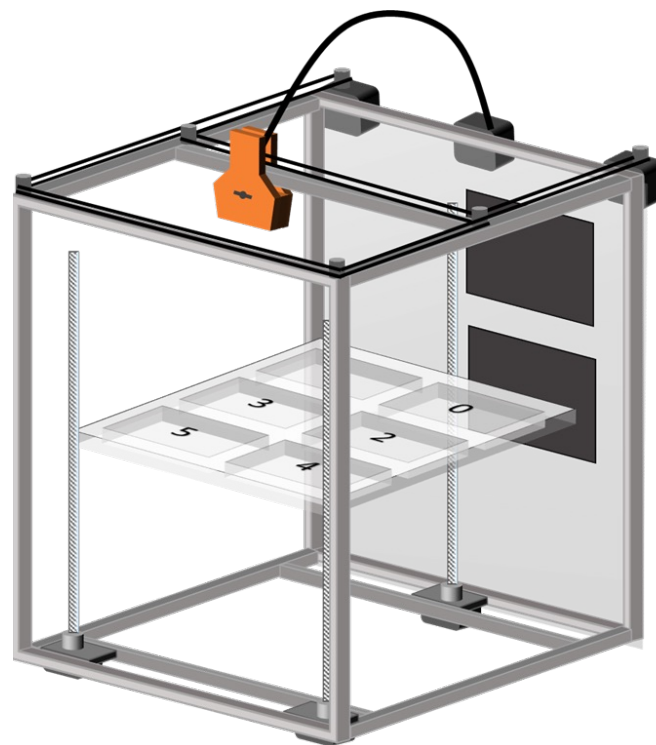
Sample Processing



Sample
Characterization



Decision Making
Algorithm



SDL demonstration: color-matching campaign

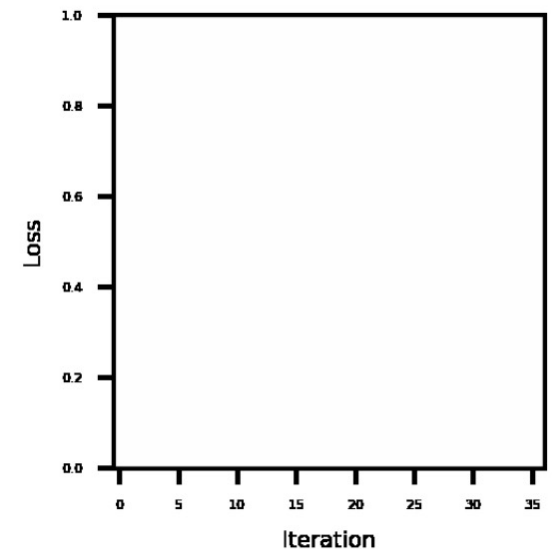


Target Color:  Train Data : 6 , Iterations: 30

Most Recent Photo



Color Loss Plot



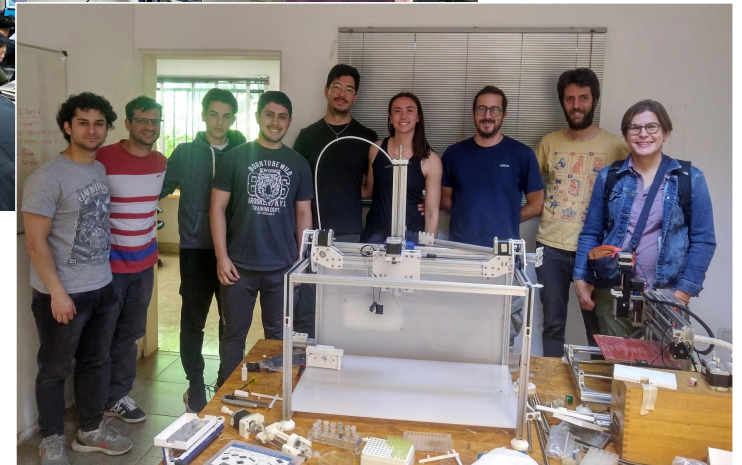
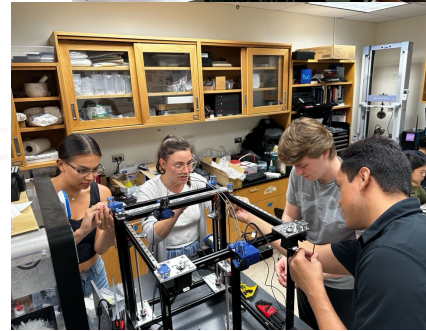
Increasing Access and Education on SDLs Across the World



Clubes de Ciencia,
Guadalajara,
Mexico

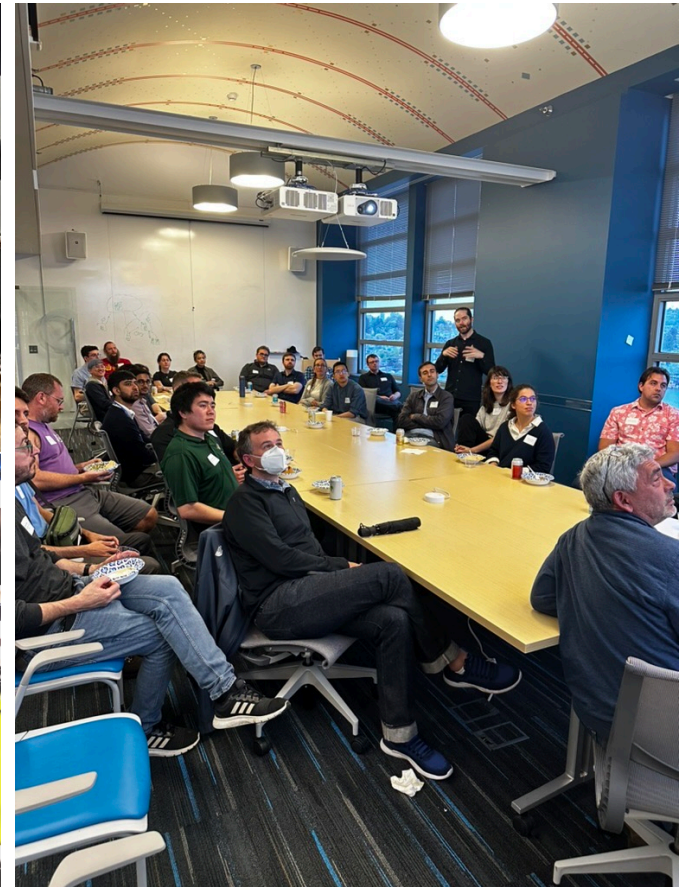
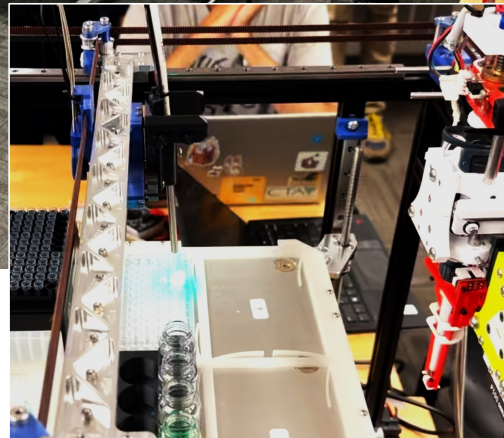
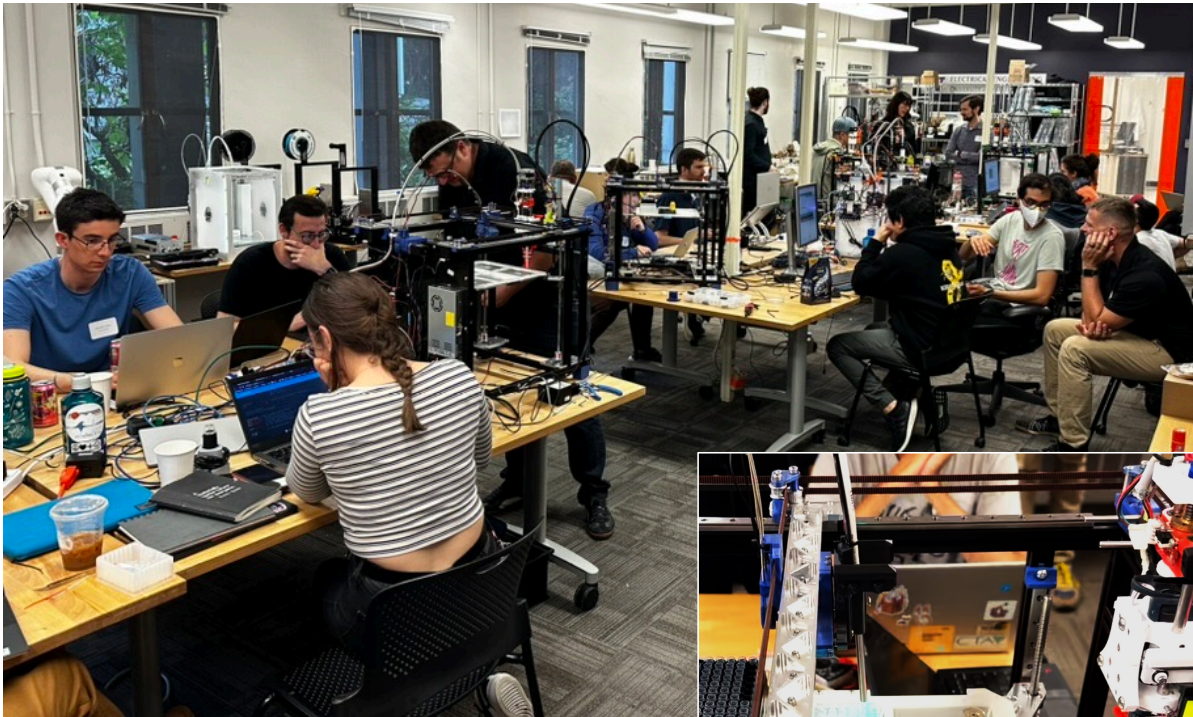


UNIVERSITY
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MĀNOA



UNSAM Buenos Aires Argentina
Instituto de Nanosistemas
Gaston Corthei - Pipetin

Building Community Across the World

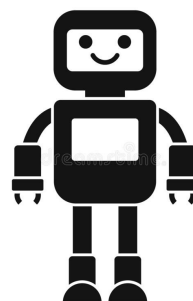
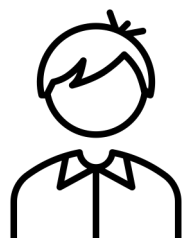


POSE Workshop
Seattle WA, April 2024

AI-Driven Retrosynthesis

Retrosynthesis: Solve Inverse Design Problems by Iterative Learning

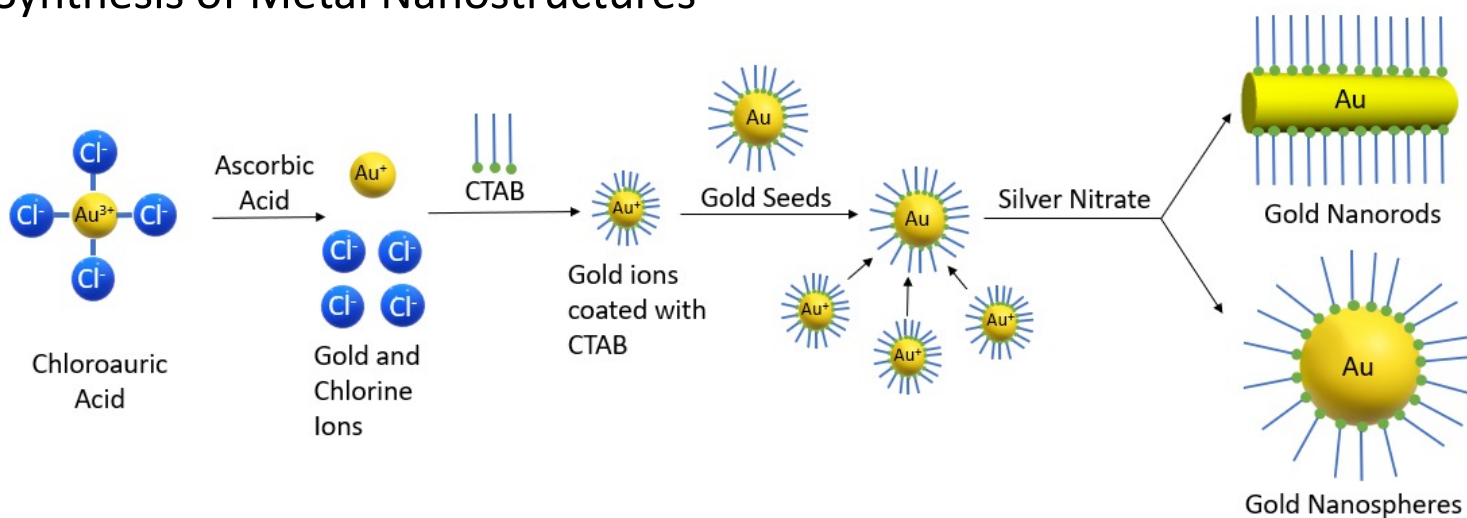
Can you help me make monodisperse rod particles of 5 nm radius and 40 nm length?



I think I can!
But let me learn it first

'Fruit Fly' Problems: Synthesis of Metal Nanostructures

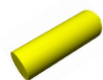
- Metal salt(s)
- Reducing Agent
- Surfactant (CTAB)
- Seeds
- Order of addition
- T, P



Automated Retrosynthesis

1. Pick Target

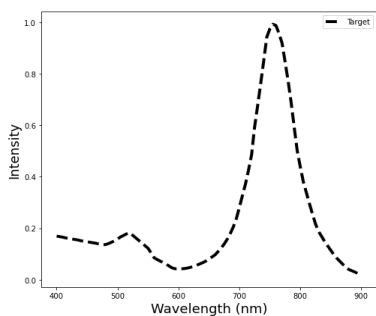
Gold Nanorod of 55nm in Length and 10nm in diameter



Spectra Scattering...

Forward Model

2. Simulate Target Data

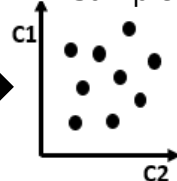


3. Perform Optimization

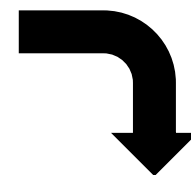
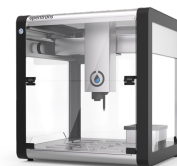
Bayesian Optimization



Composition of Samples



Robotic Synthesis



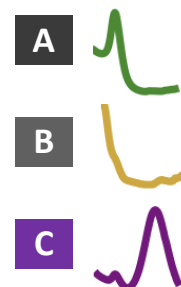
Sample A

Sample B

Sample C



Characterization



Distance Calculation

Target



Score

Comp. A

2

Comp. B

3

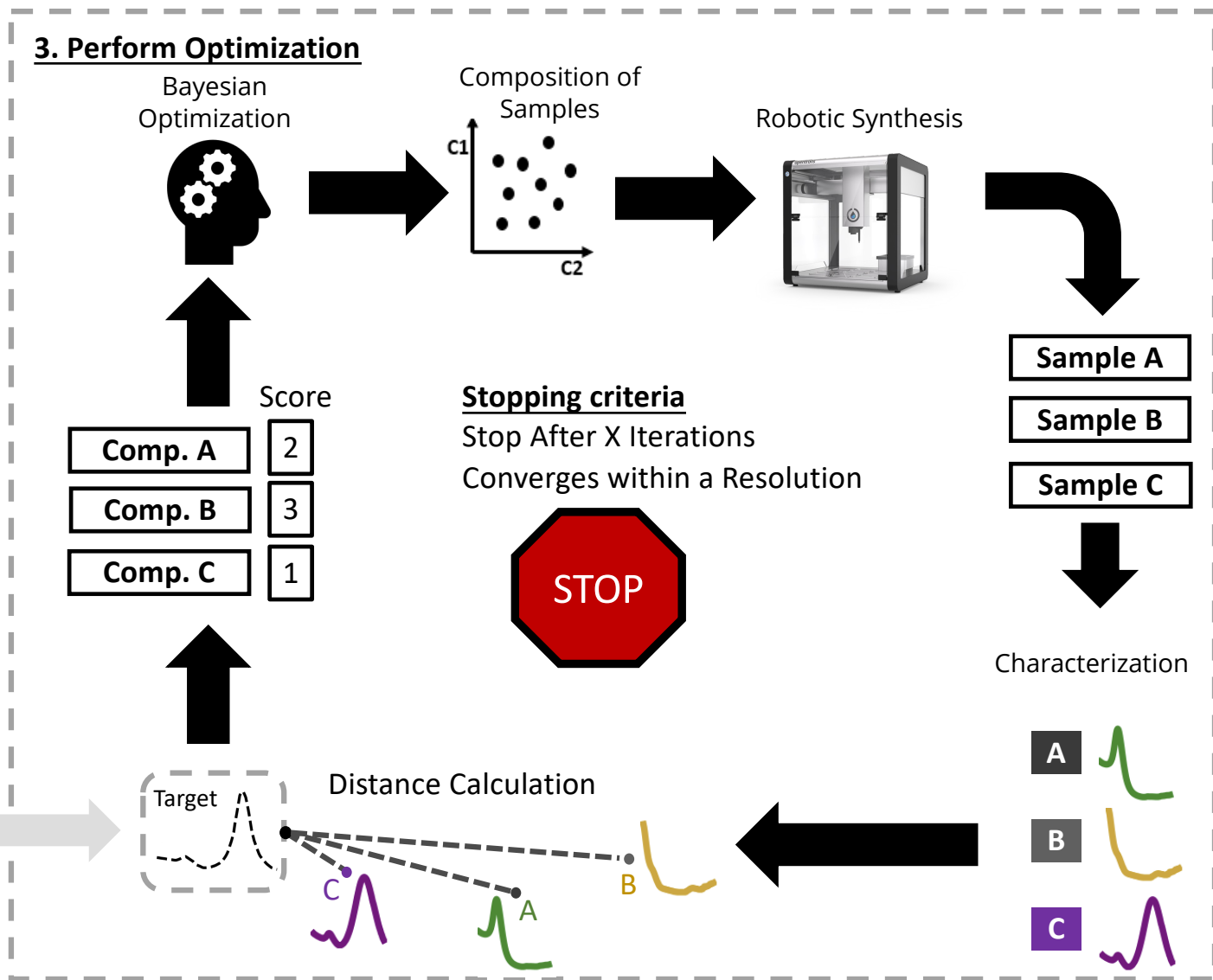
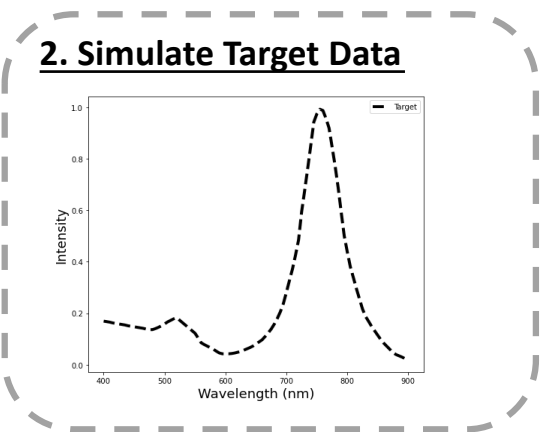
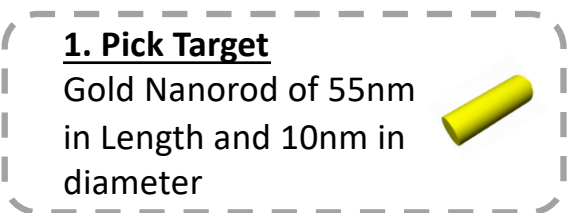
Comp. C

1

Stopping criteria

Stop After X Iterations

Converges within a Resolution



Cold, warm, warmer, hot... search game

Distance Metrics

LIKELIHOOD

The probability of "B" being True, given "A" is True

PRIOR

The probability "A" being True. This is the knowledge.

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$

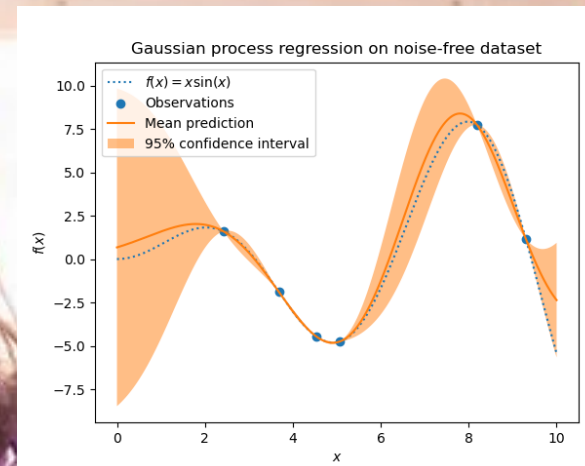
POSTERIOR

The probability of "A" being True, given "B" is True

MARGINALIZATION

The probability "B" being True.

Bayesian Optimization



GP Regression

How to Meaningfully Compare Trials to Target?

- **Match in 'Expert' Parameters**

- Peak Wavelength Intensity and Position (scalar values)

- **Euclidean Distance**

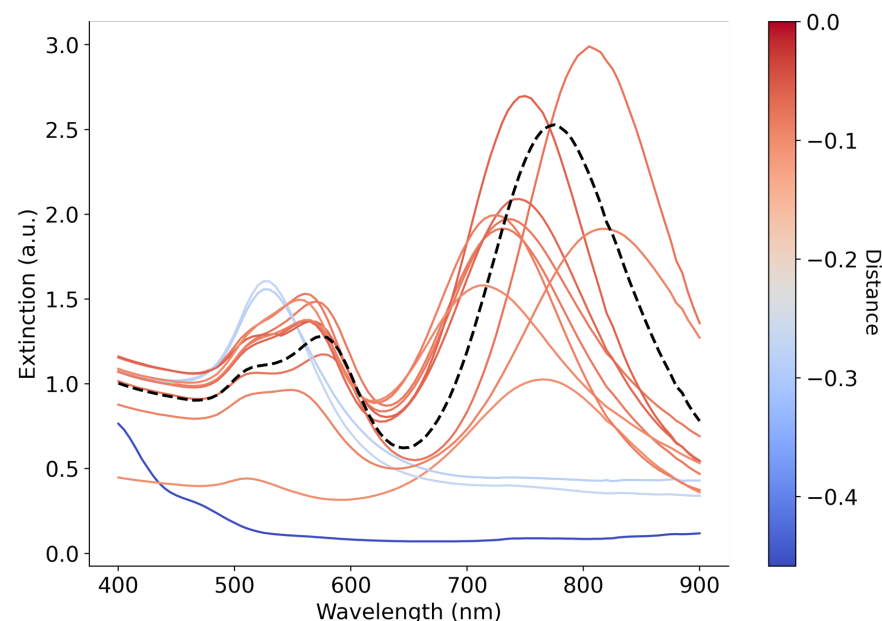
- Mean squared difference of two spectra

- **Square Root Slope Function (SRSF)**

- Compares the shape of spectra by taking derivatives

- **Amplitude-Phase Distance**

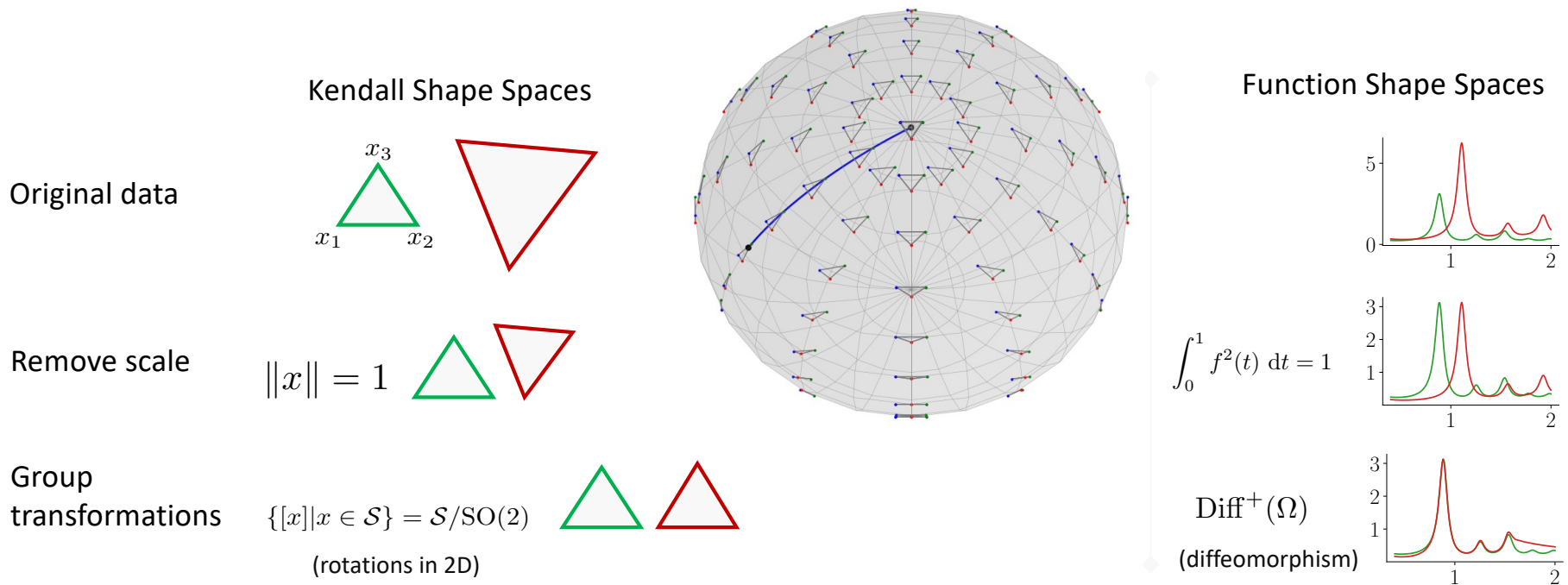
- Accounts for variation along x-axis and y-axis after 'alignment' in function space



“Autonomous retrosynthesis of gold nanoparticles via spectral shape matching”

K. Vaddi, H. Chiang, L. Pozzo, Digital Discovery, 2022

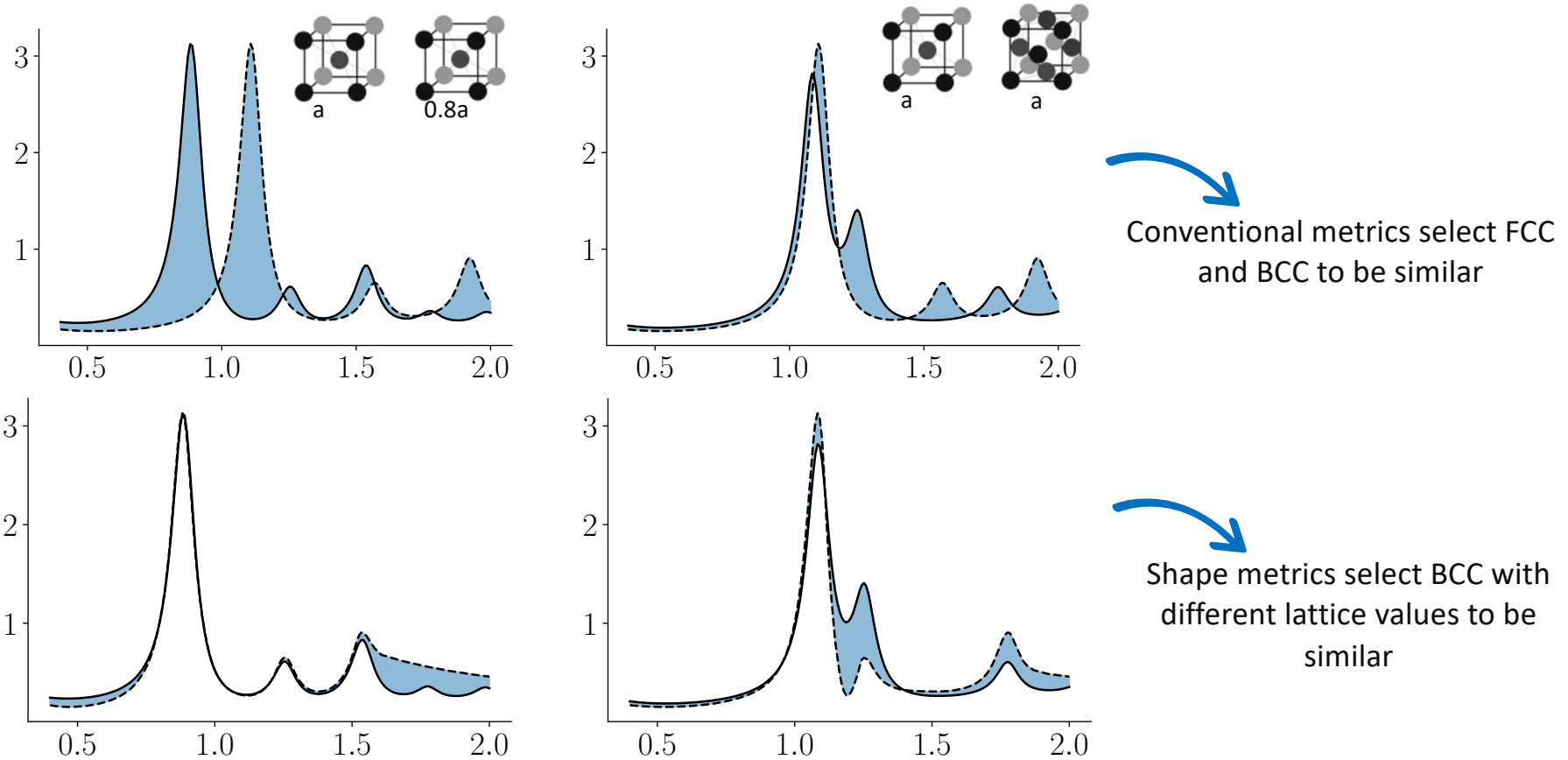
'Distance' in Function Shape Spaces



KENDALL, David G. et KENDALL, Wilfrid S. Alignments in two-dimensional random sets of points. Advances in Applied probability, 1980, p. 380-424.

<https://geomstats.github.io>

Shape-Based Similarity in Phase Identification



“Metric geometry tools for automatic structure phase map generation”, K. Vaddi, K. Li, L. Pozzo, *Digital Discovery*, 2, 1471, (2023)

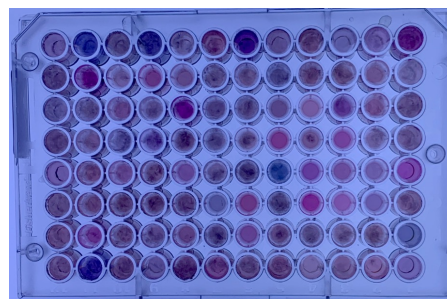
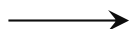
Seed-mediated growth of gold nanorods

Reagent	Stock (M)	Target (M)	Range (M)
CTAB	2.0×10^{-1}	6.40×10^{-2}	6.40×10^{-2}
Gold (III) Chloride Trihydrate	1.0×10^{-3}	1.96×10^{-4}	1.96×10^{-4}
Silver Nitrate	6.4×10^{-4}	6.20×10^{-5}	$0 - 7.38 \times 10^{-5}$
Ascorbic Acid	6.3×10^{-3}	3.60×10^{-4}	$0 - 7.27 \times 10^{-4}$
Gold Seeds	1.8×10^{-5}	1.44×10^{-6}	1.44×10^{-6}

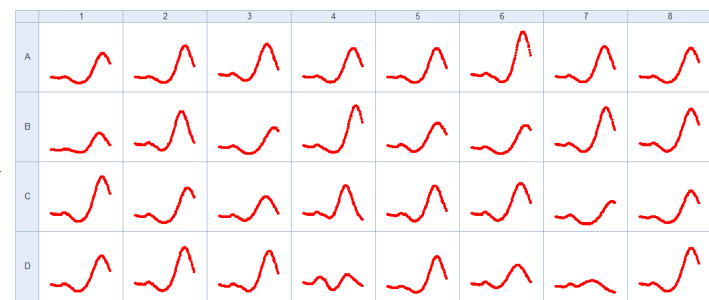
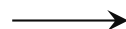
Two-dimensional
design space



Sample preparation



Synthesized samples



UV-Vis characterization

“Autonomous retrosynthesis of gold nanoparticles via spectral shape matching”

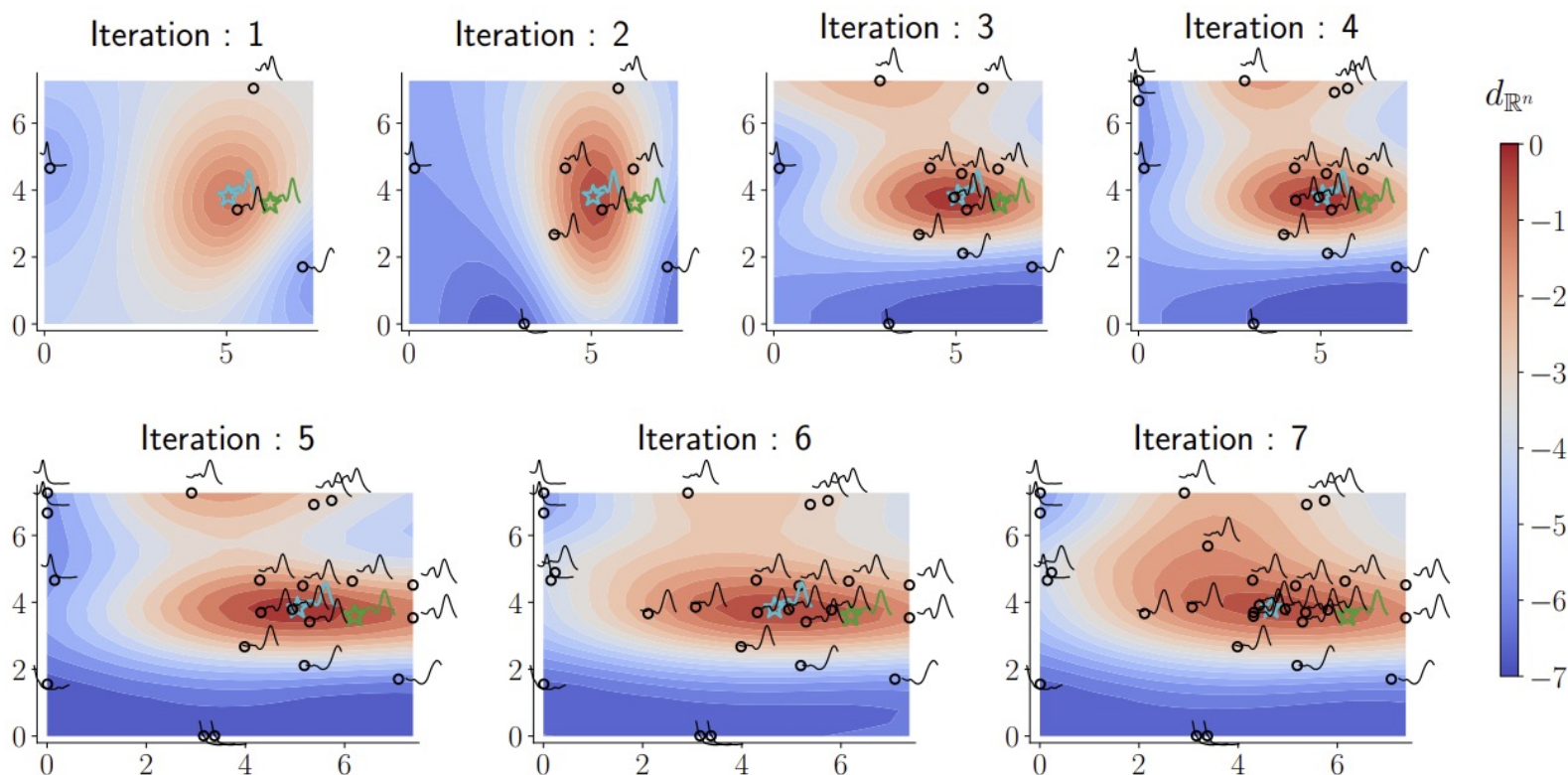
K. Vaddi, H. Chiang, L. Pozzo, Digital Discovery, 2022

Amplitude Phase Metric Helps Agent Learn the Phase Diagram

Euclidean Metric

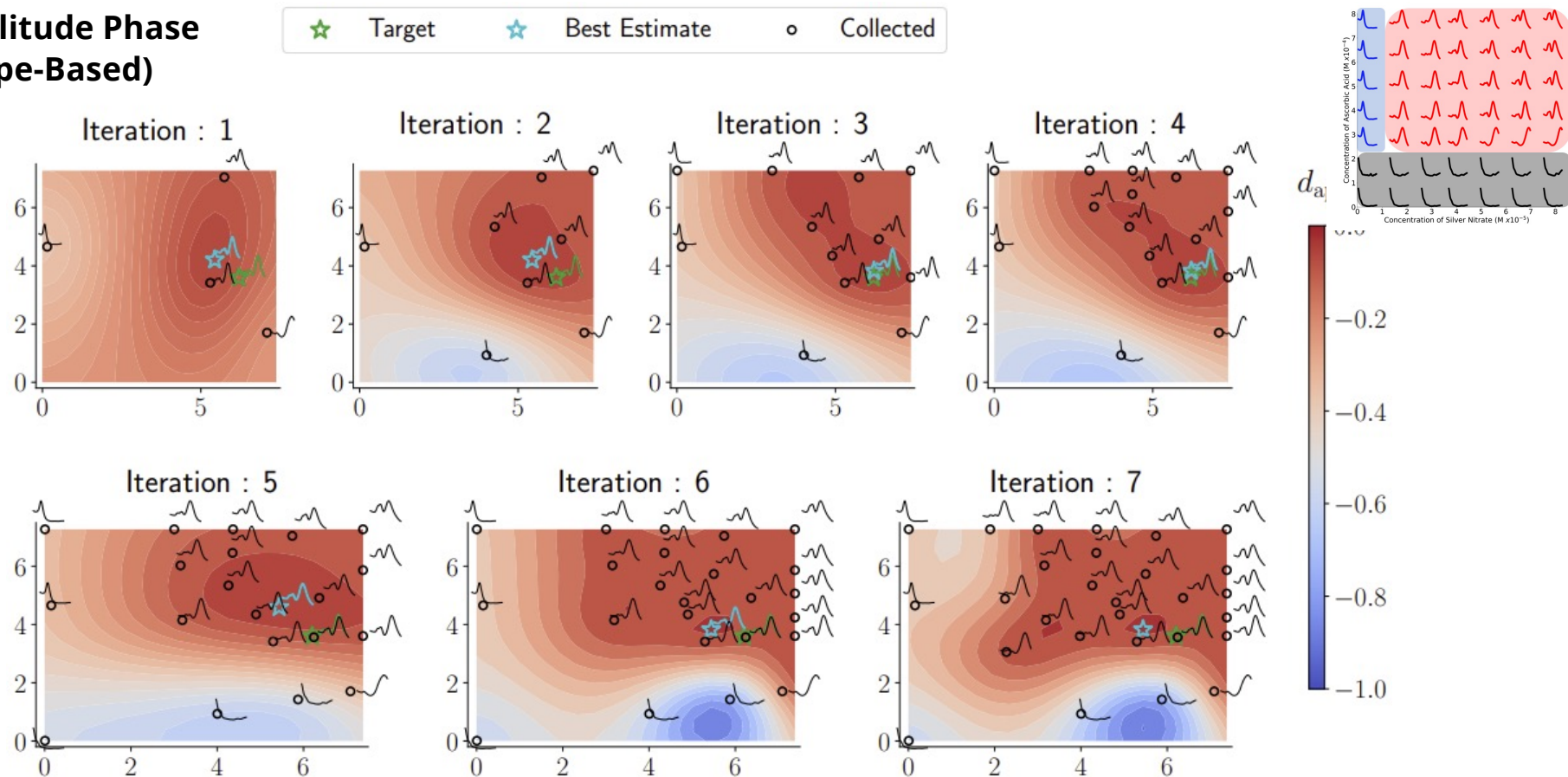
★ Target ★ Best Estimate ○ Collected

Bayesian Optimization with GP Regressor



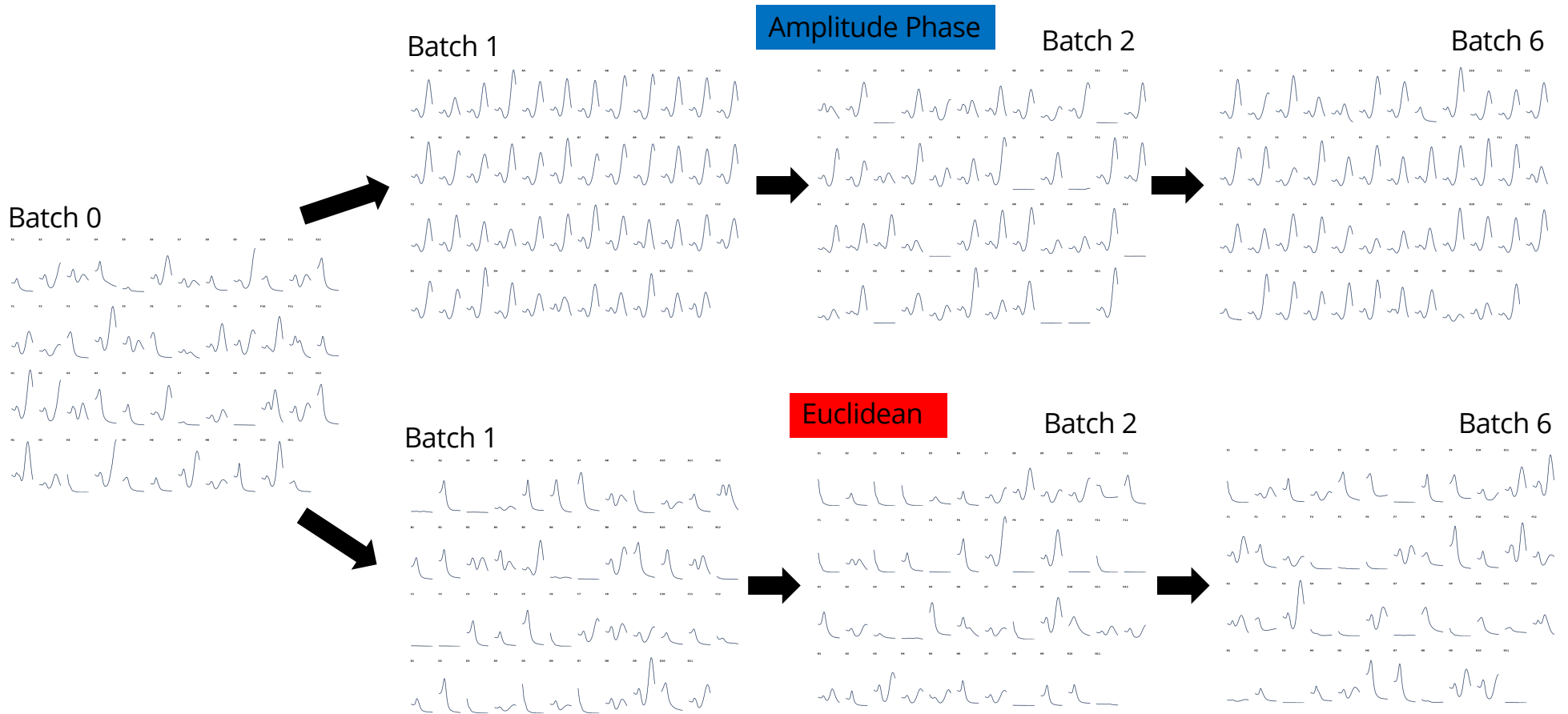
Amplitude Phase Metric Helps Agent Learn the Phase Diagram

Amplitude Phase (Shape-Based)

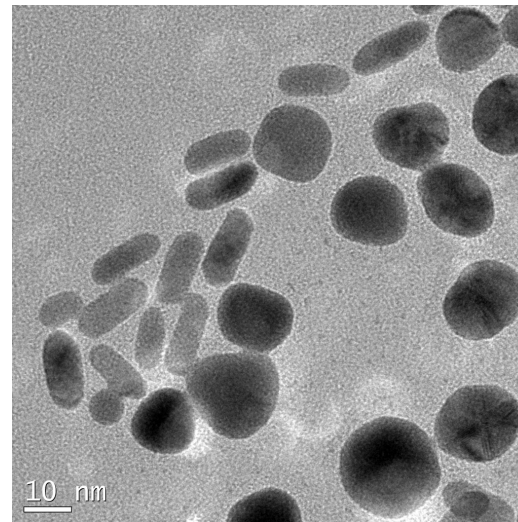
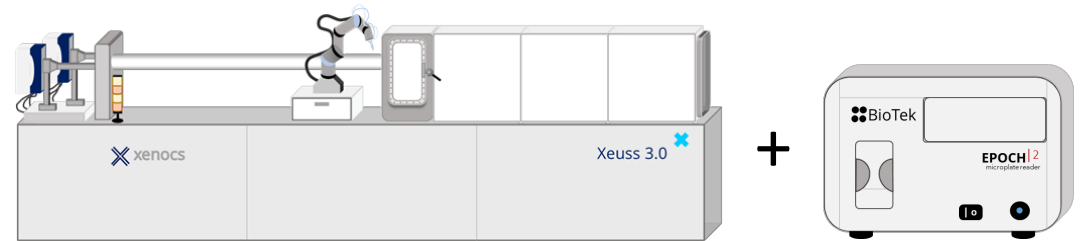
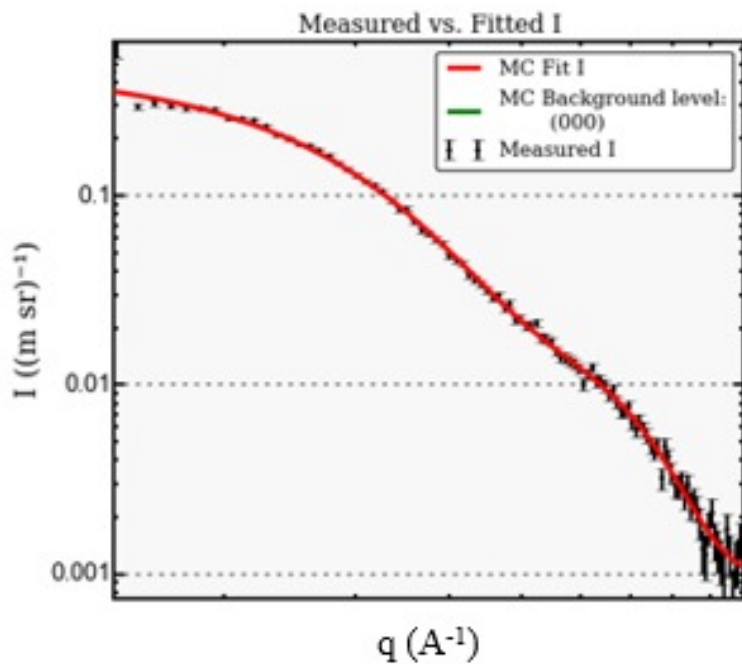


“Autonomous retrosynthesis of gold nanoparticles via spectral shape matching” K. Vaddi, H. Chiang, L. Pozzo, Digital Discovery, 2022

Amplitude Phase Metric Helps AI Identify Nano-Rod Space



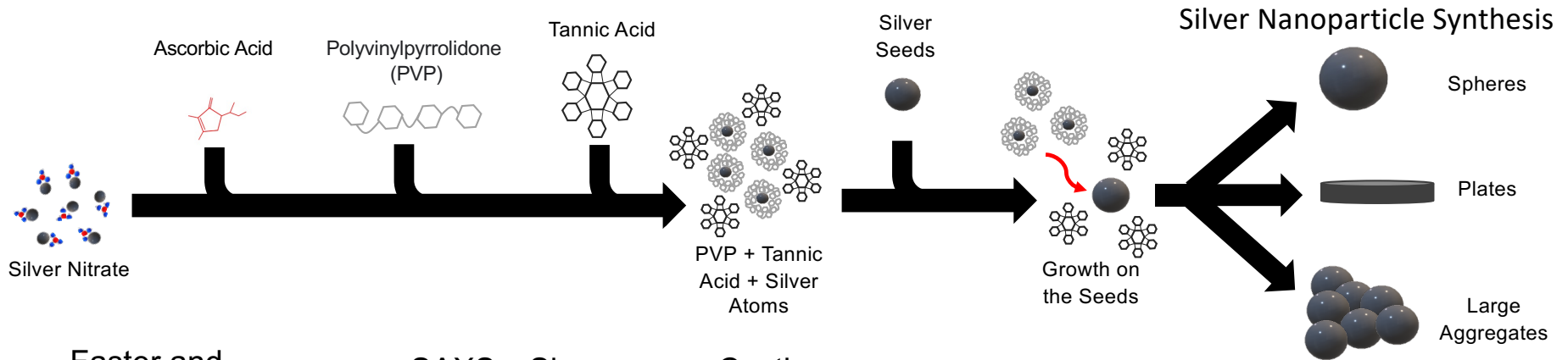
Did it work out?... Not as Expected



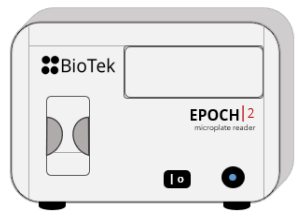
SAXS: nanorods of 8 nm in diameter and 15 nm in length and nanospheres of 16 nm in diameter

Need Multi-Fidelity Models!!

Hierarchical Multifidelity Analysis of Silver Nanoplates

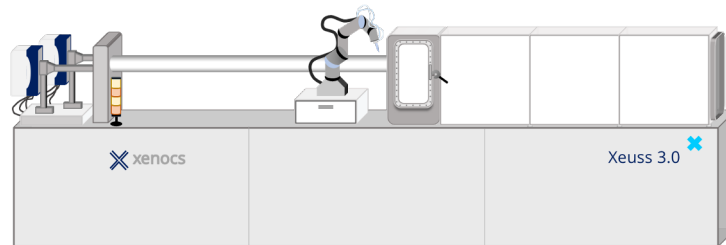


Faster and Cheaper



UV-Vis Spectroscopy

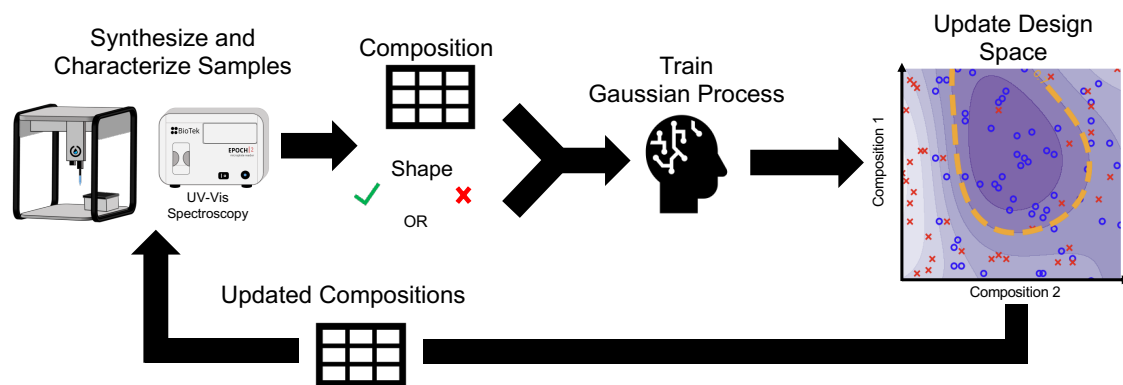
SAXS Slower more Costly



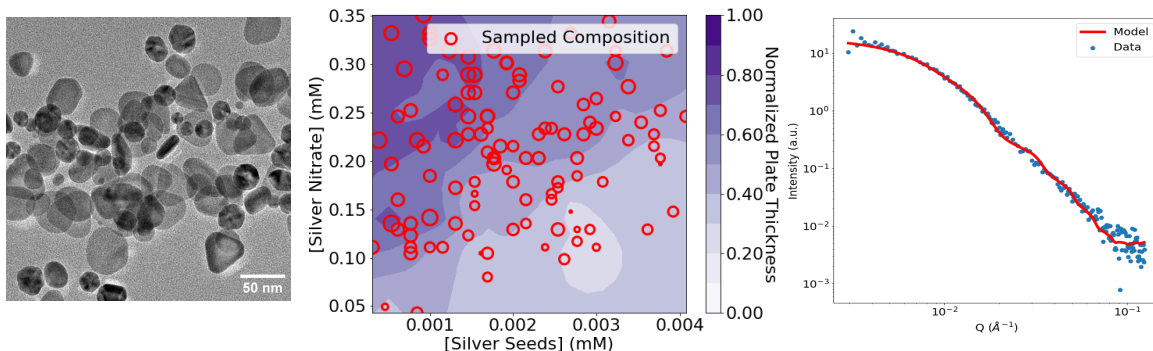
TEM Slowest and Costliest



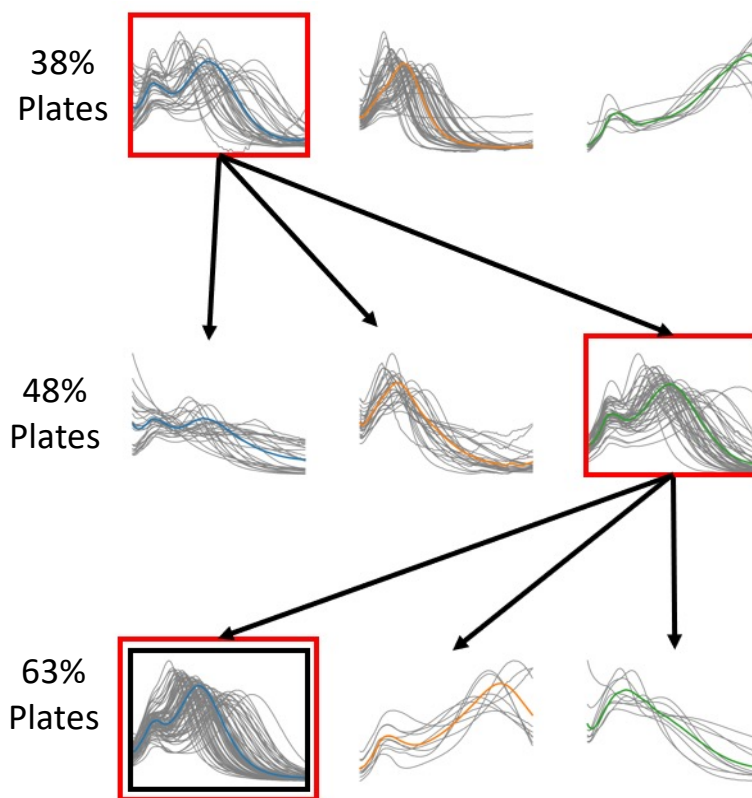
Hierarchical and Multi-Modal Analysis



Optimized sampling to observe plate formation



Autonomous Classification of Spectroscopic Curves Based on Shape



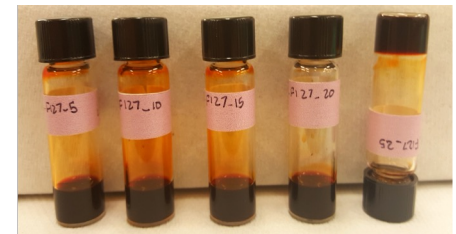
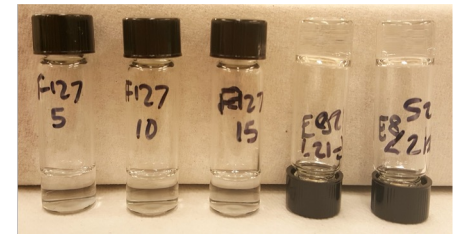
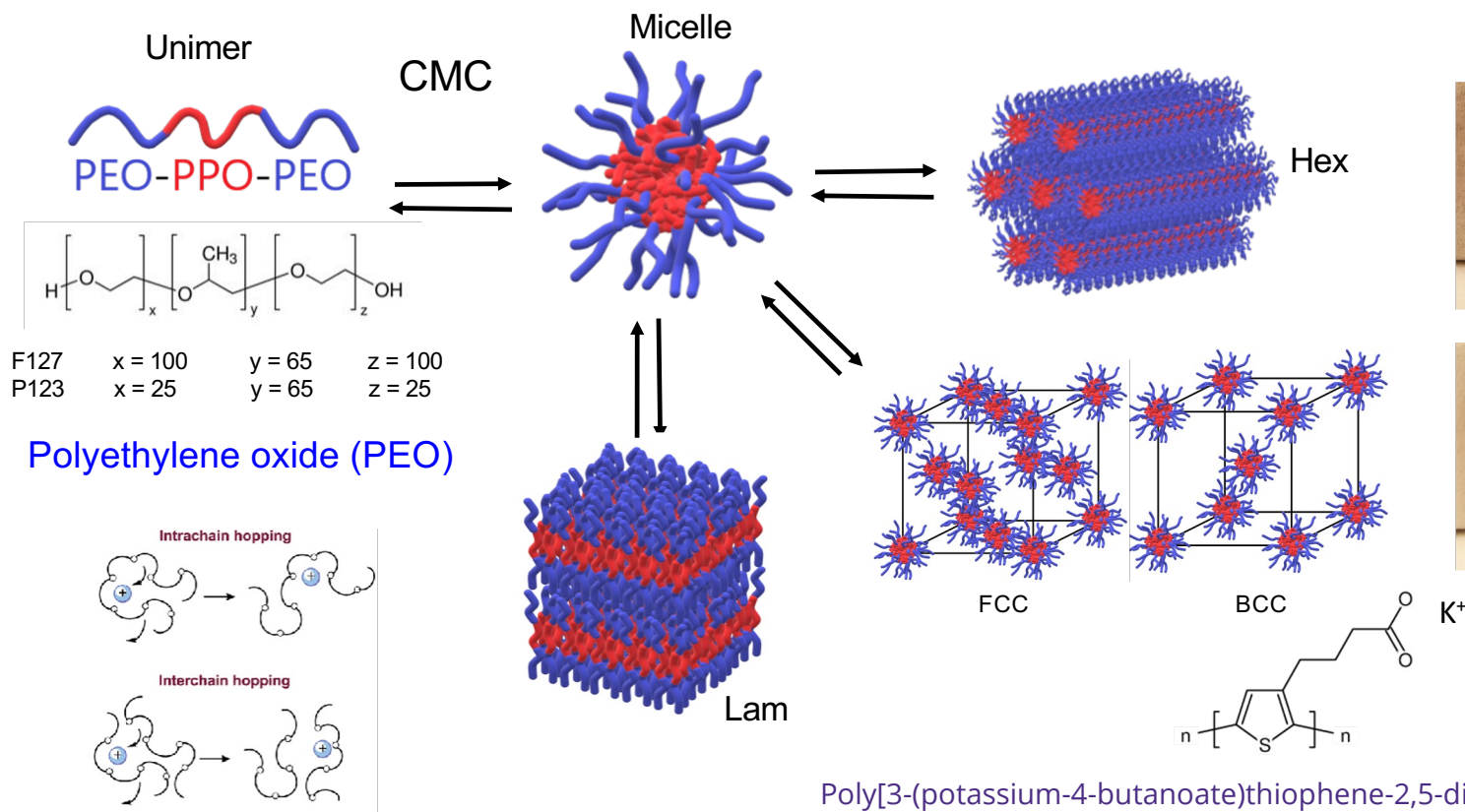
Autonomous Phase Mapping

Integration of automation and high-throughput analysis for SANS and SAXS

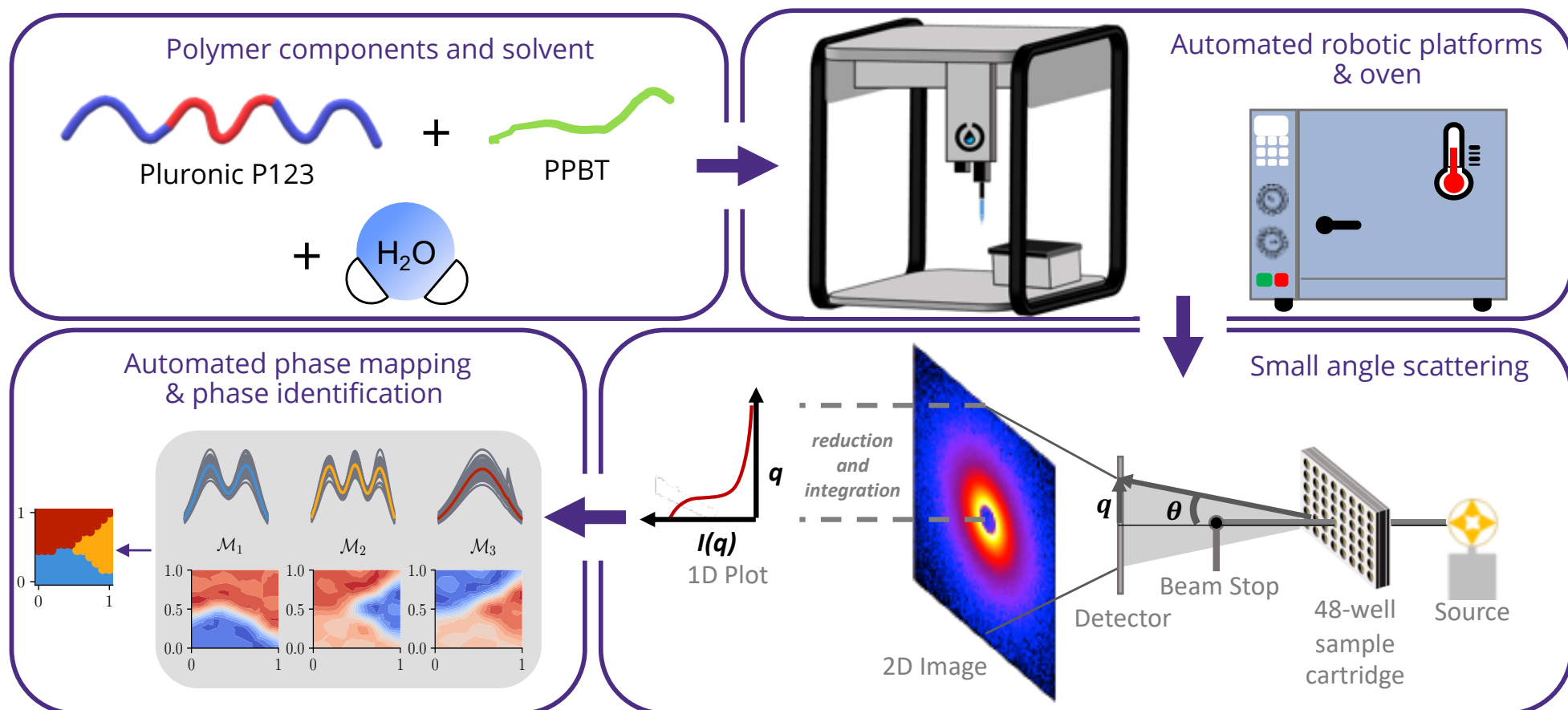


<https://github.com/pozzo-research-group/Automation-Hardware>

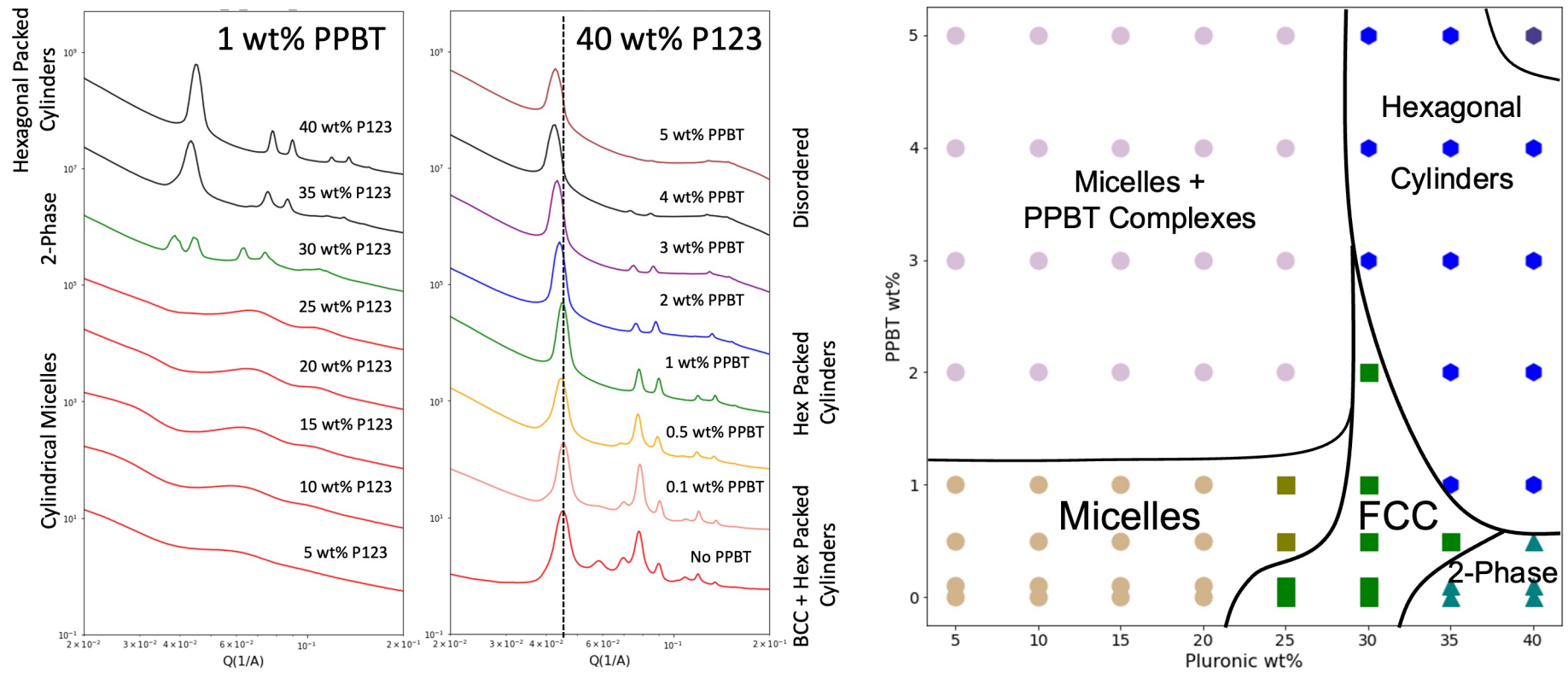
Phase-Mapping: Conjugated Polymer + Copolymer



Automated High-Throughput OMIEC Formulation and Analysis



HT SAXS Mapping of Structured Blend Phase-Diagram

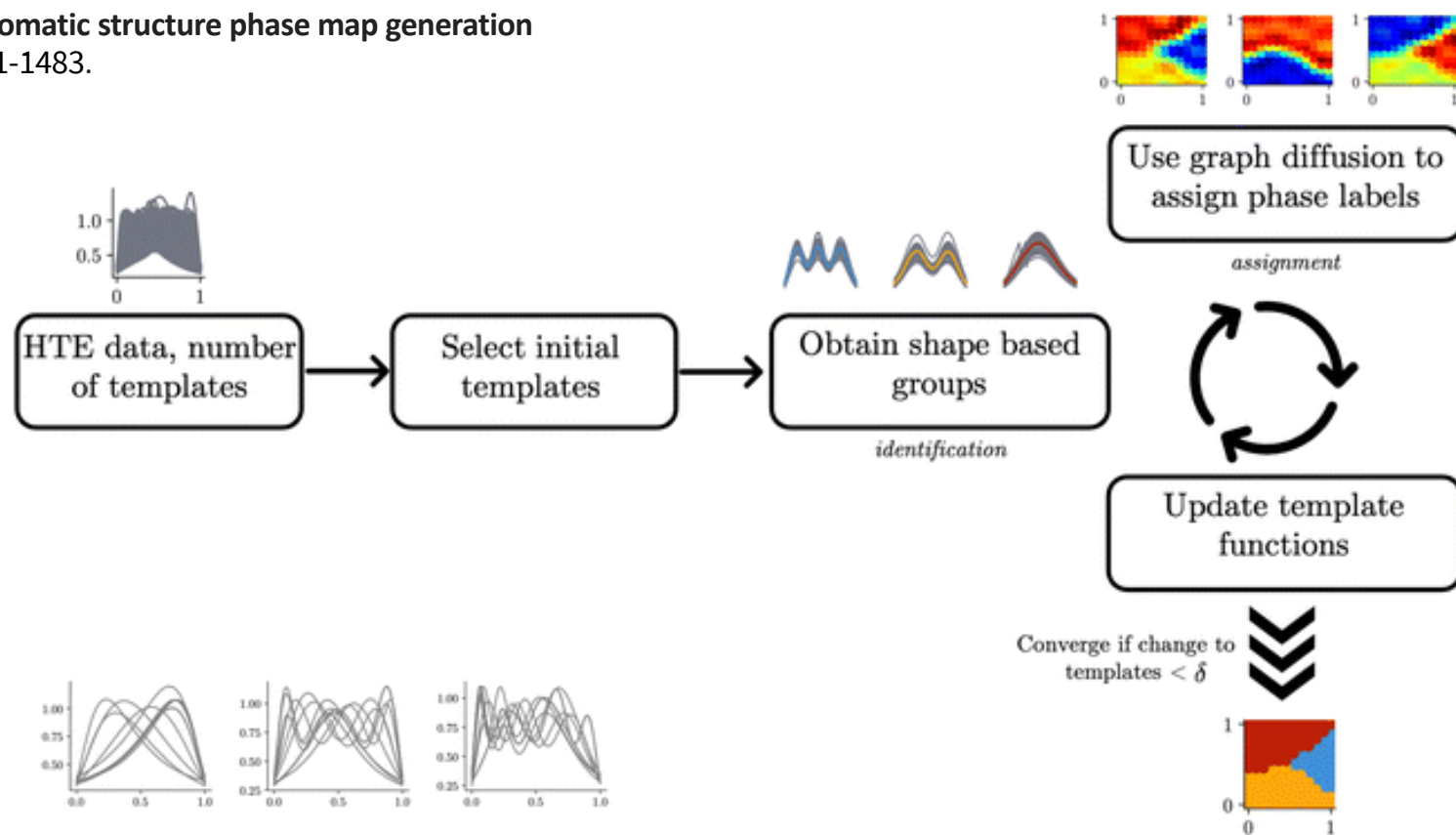


Autonomous Phase-Mapping with Shape-Similarity Distance Metrics

Kiran Vaddi, Karen Li, and Lilo D. Pozzo.

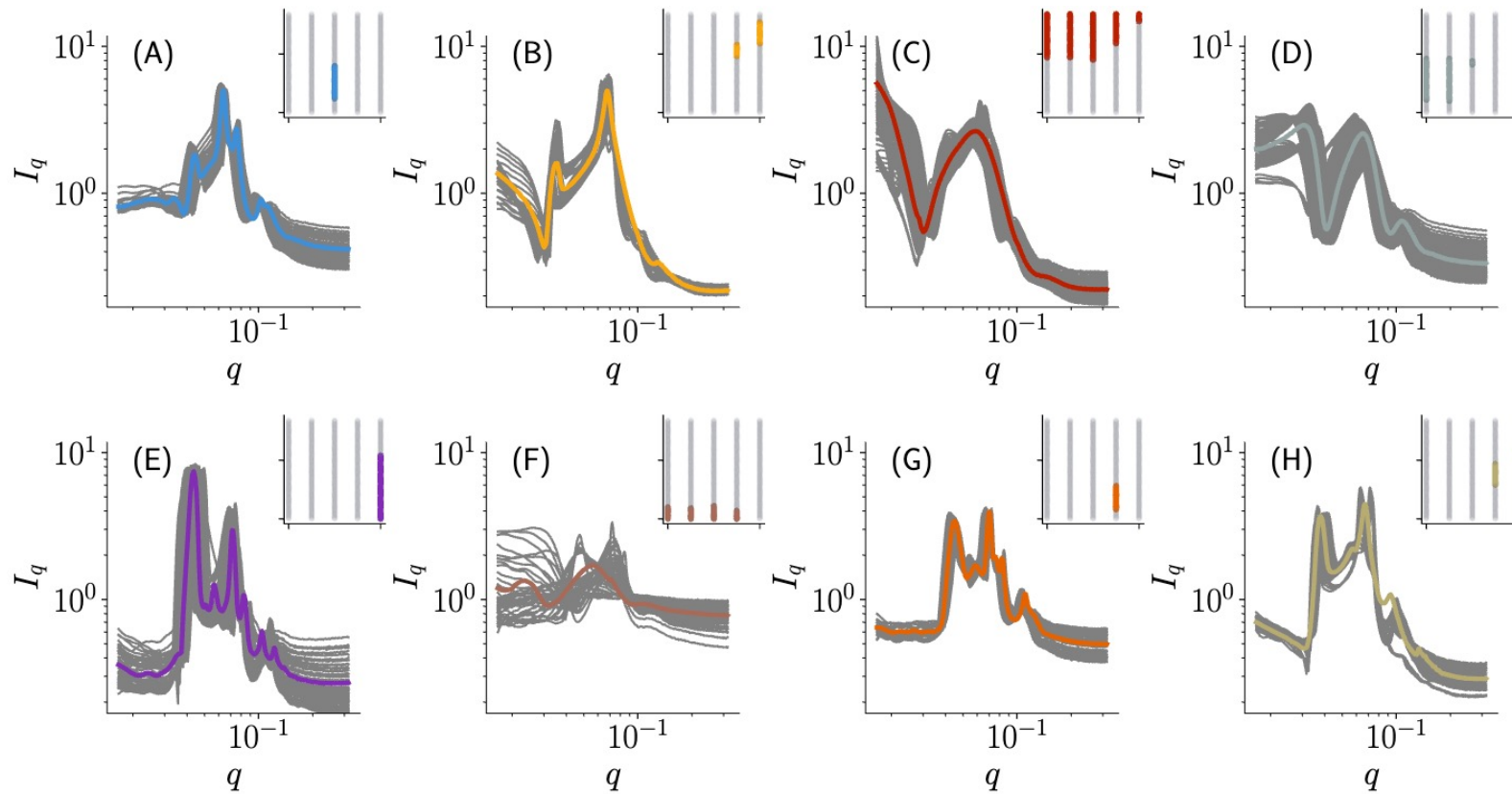
Metric geometry tools for automatic structure phase map generation

Digital Discovery, 2023,2, 1471-1483.



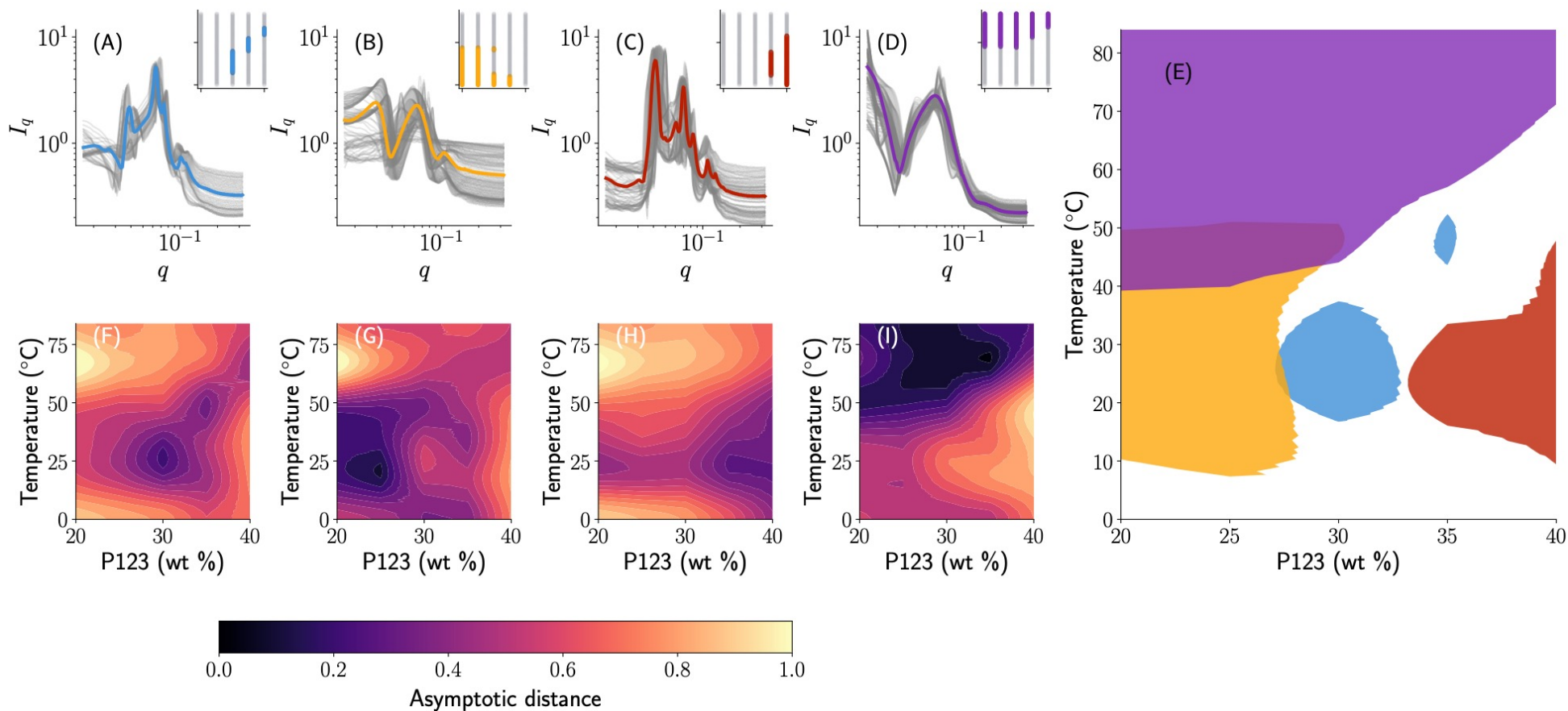
Code GitHub Repo

Distance Metrics in Autonomous Phase Mapping

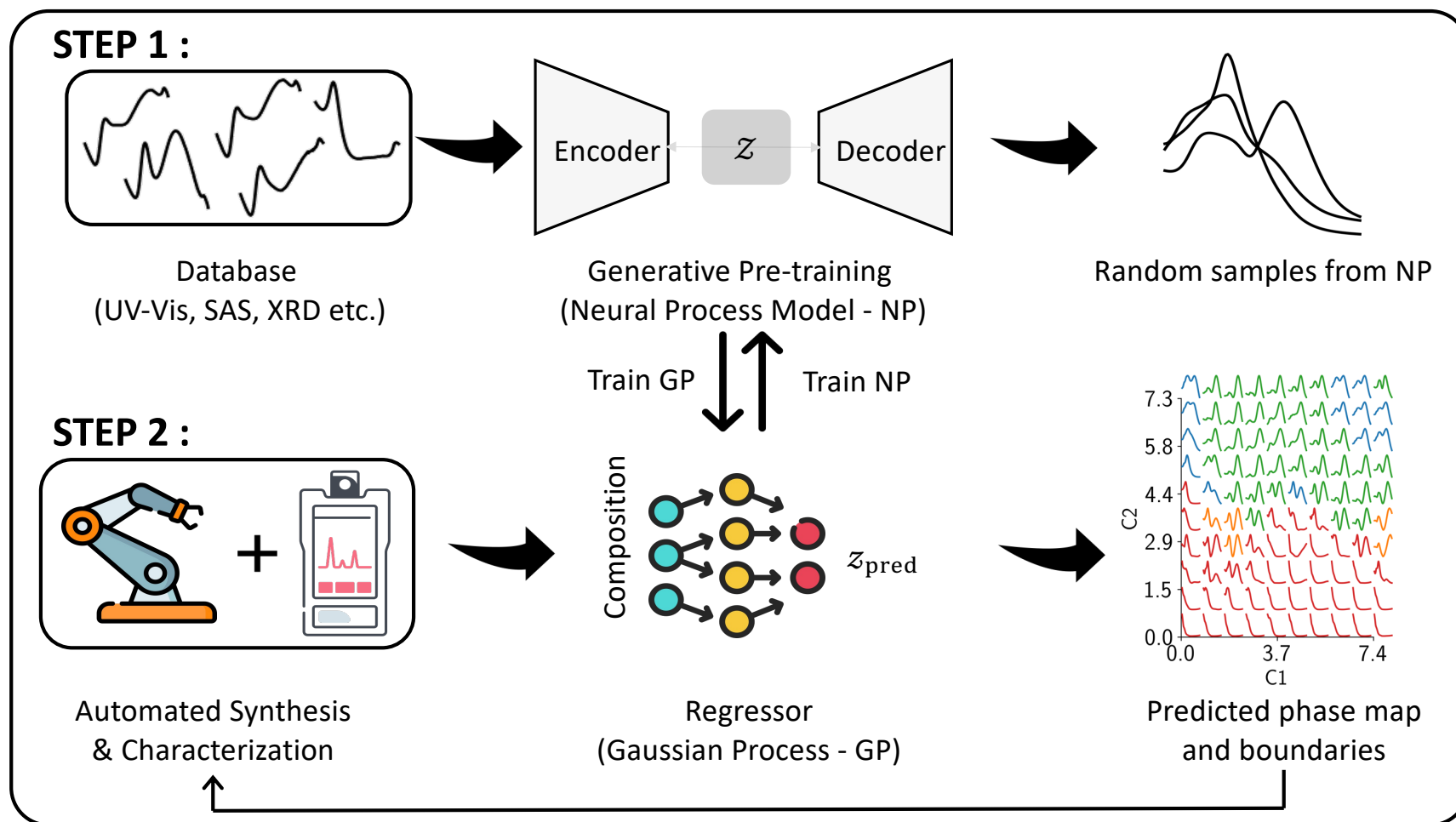


“Metric geometry tools for automatic structure phase map generation”, K. Vaddi², K. Li², L. Pozzo, Digital Discovery, 2, 1471-1483, (2023)

Distance Metrics in Autonomous Phase Mapping

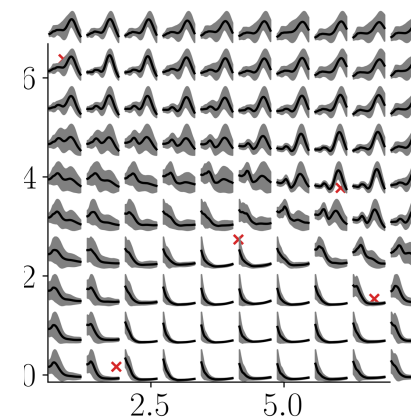
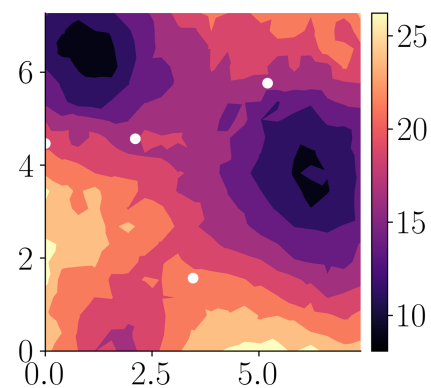
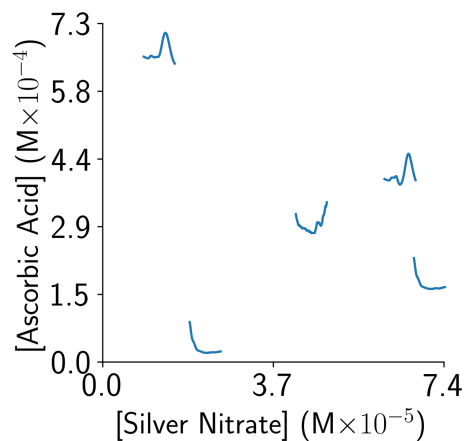


Two-step workflow for “differentiable” phase mapping

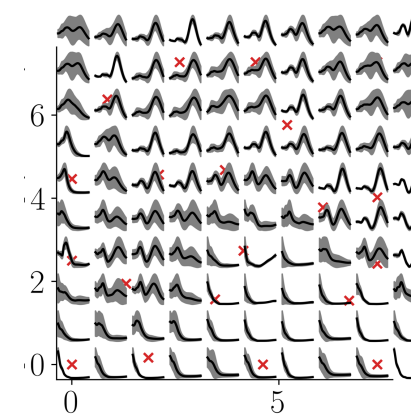
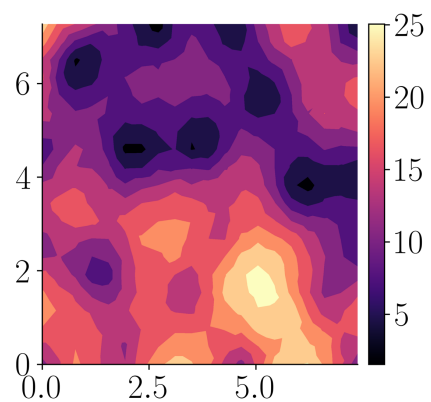
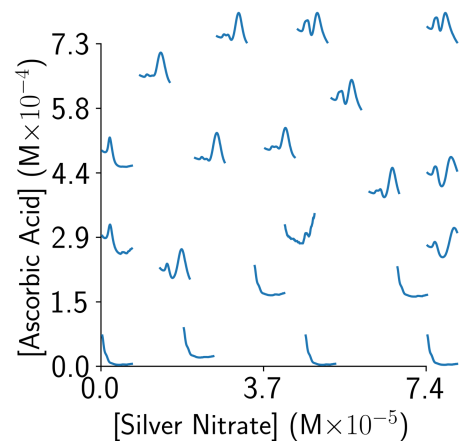


Learning to predict spectra from experiments

Iteration 1



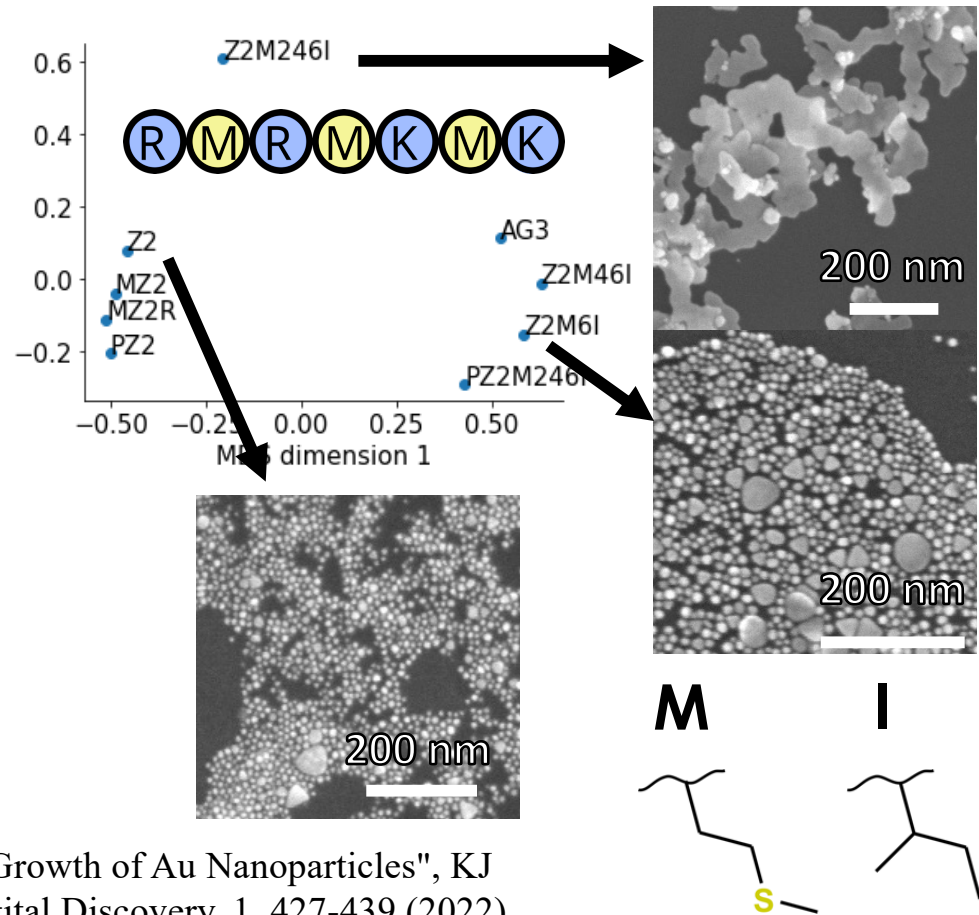
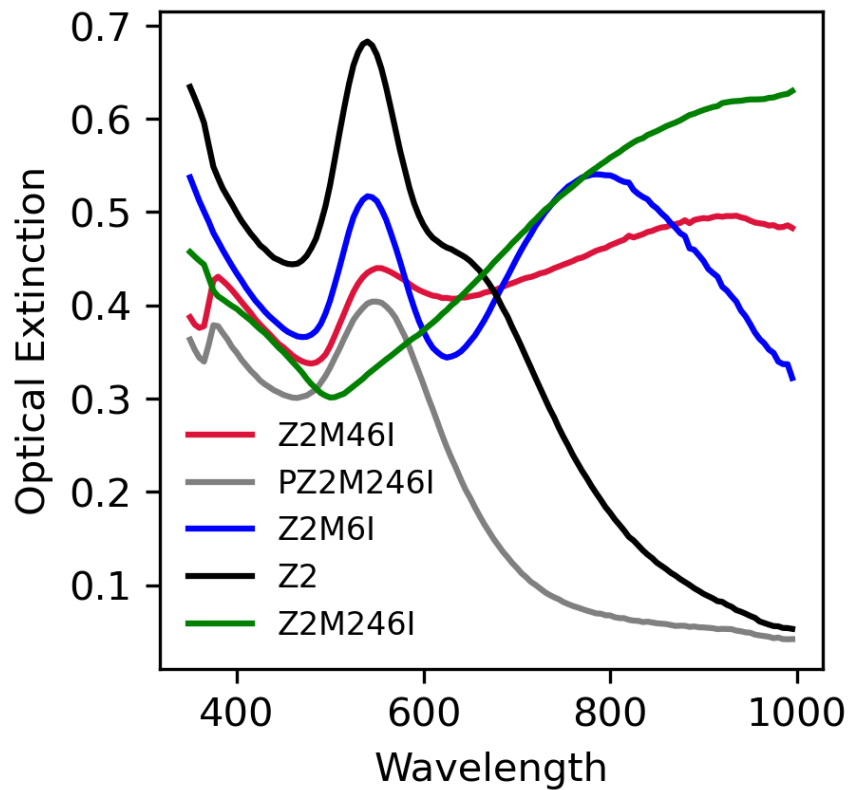
Iteration 5



Timed Intervention

States Out-of-Equilibrium

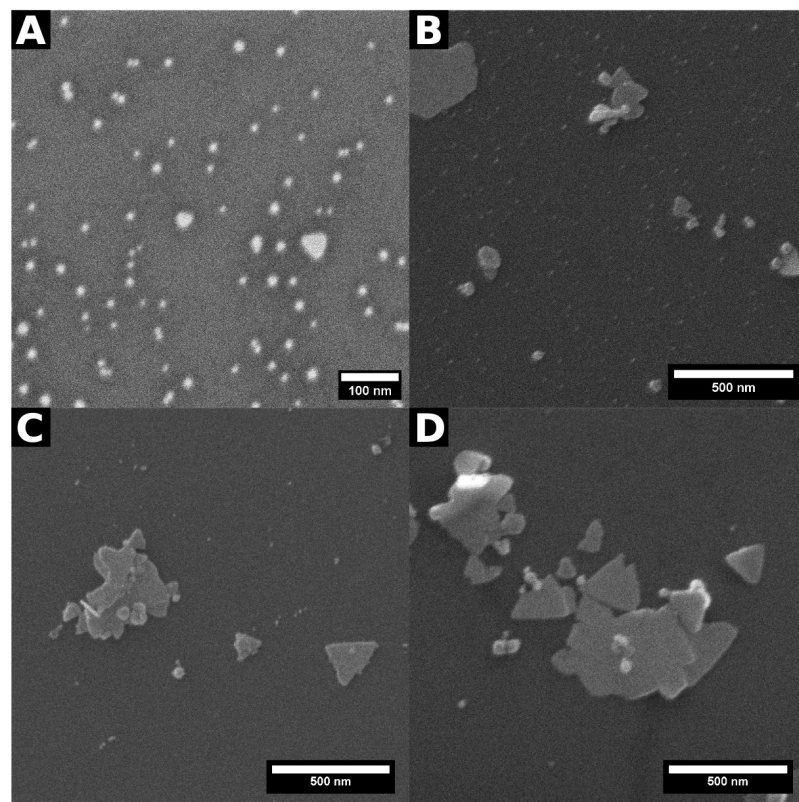
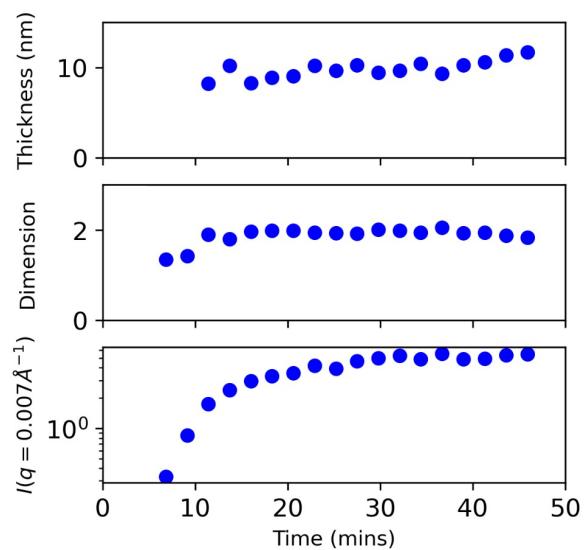
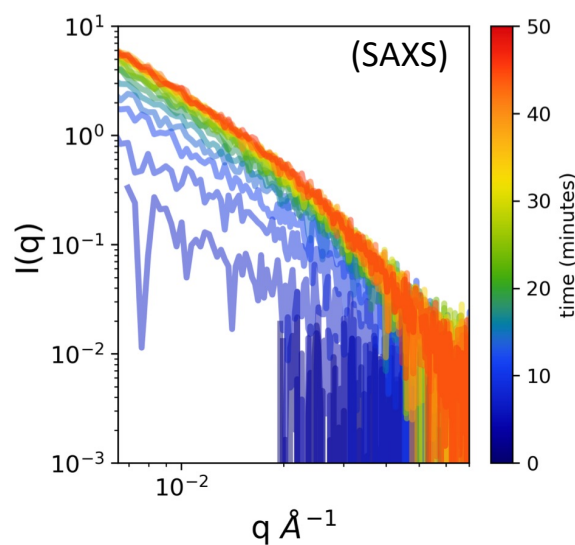
Dynamic Intervention: Peptide-Mediated Growth



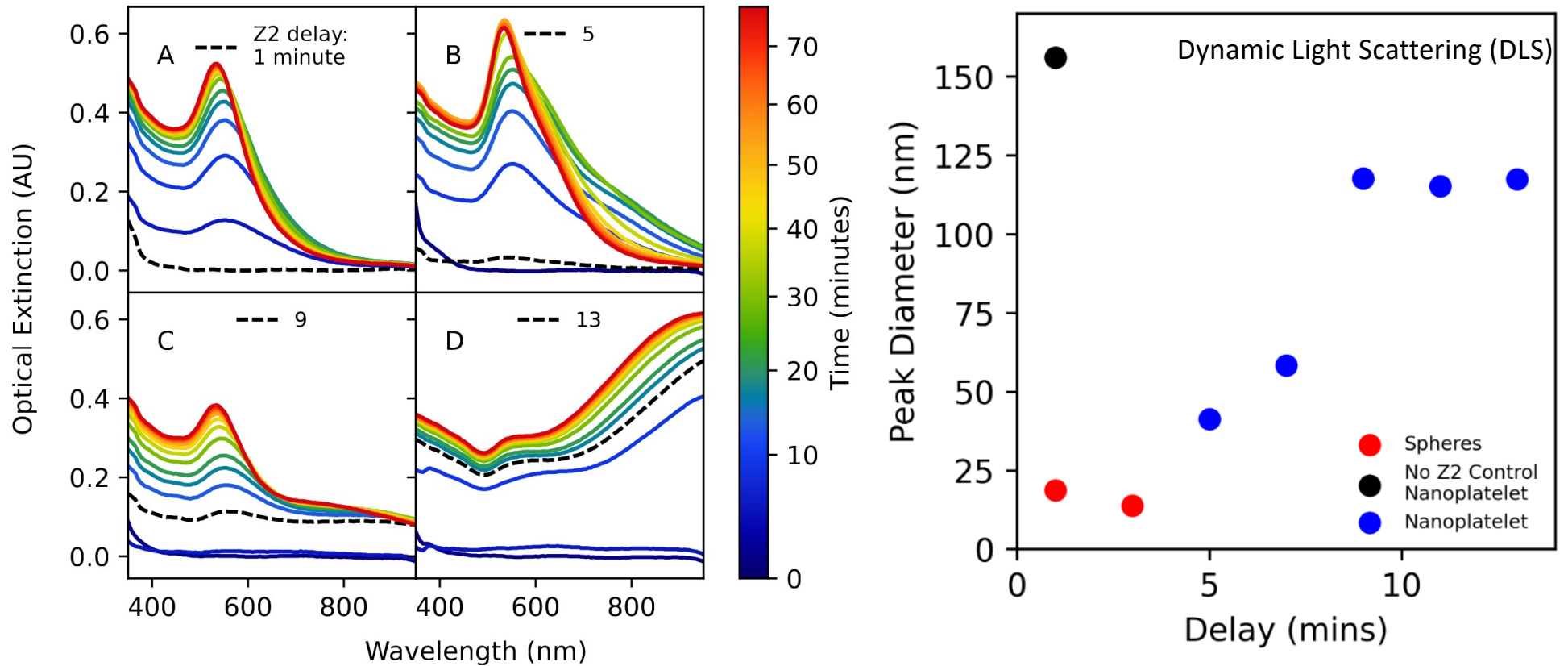
"Multivariate Analysis of Peptide-Driven Nucleation and Growth of Au Nanoparticles", KJ Lachowski, K Vaddi, NY Naser, F Baneyx, LD Pozzo. Digital Discovery, 1, 427-439 (2022)

Dynamic Intervention: Accessing Out of Equilibrium States

Delay (mins): 0 2 4 6 8 10 12 No Z2 Control



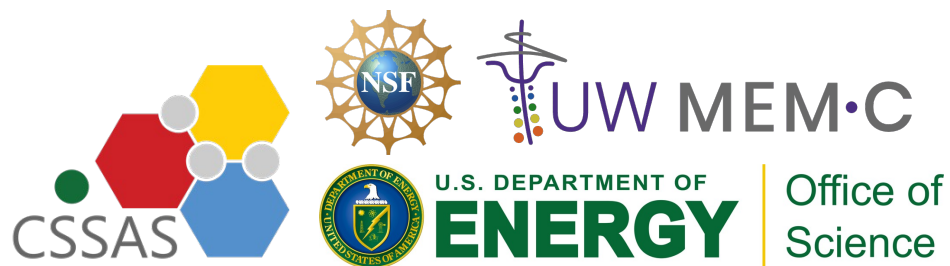
Dynamic Intervention with Z2 Stabilizes Small Plates



“Anisotropic Gold Nanomaterial Synthesis Using Peptide Facet Specificity and Timed Intervention” K. J. Lachowski, H. T. Chiang, K. Torkelson, W. Zhou, S. Zhang, J. Pfaendtner, L. D. Pozzo, Langmuir 39 (45), 15878-15888, (2023)

Conclusions

- Presented Tools for Broad Adoption of SDLs
- Distance metrics play an important role for AI agents working on functional data
- 'Shape Matching' distance (Amplitude-Phase) outperforms standard Euclidean and expert defined metrics
- Next Steps:
 - Multi-fidelity optimization with automated SAXS data
 - Pathway manipulation in non-equilibrium systems



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