

Data-centric Design of Combinatorial Materials Systems

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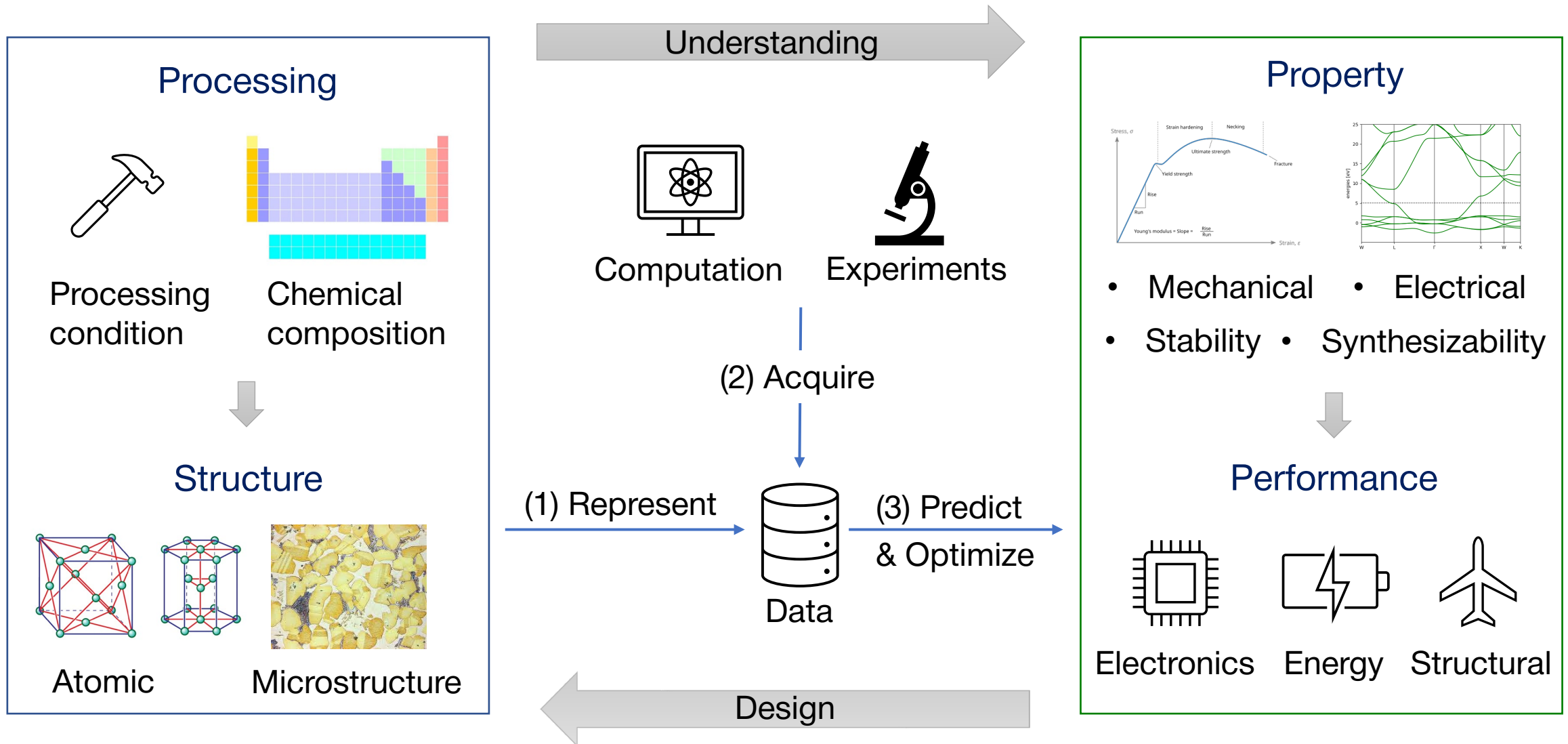


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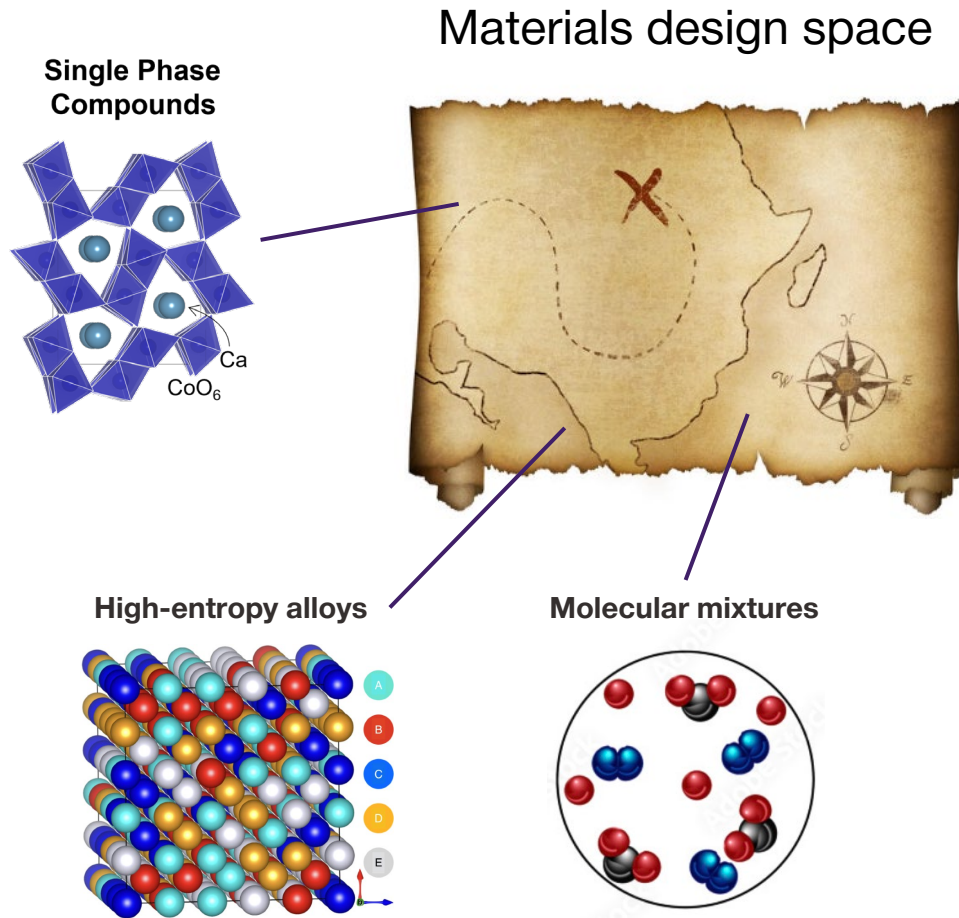


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Data-centric materials design



Structural complexity and representation



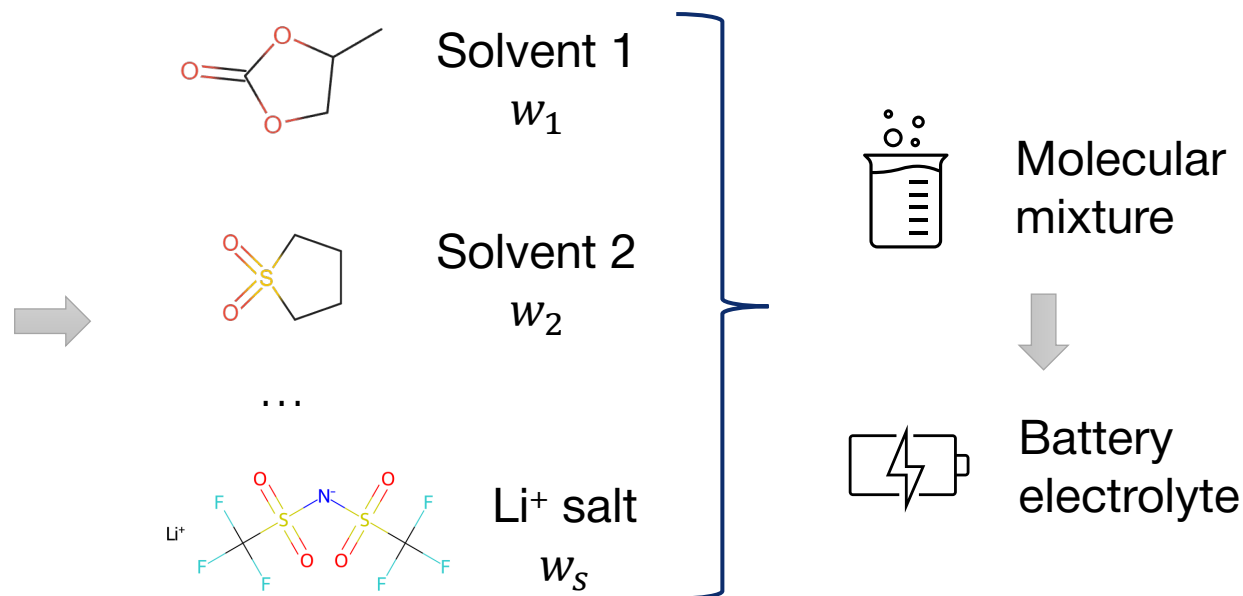
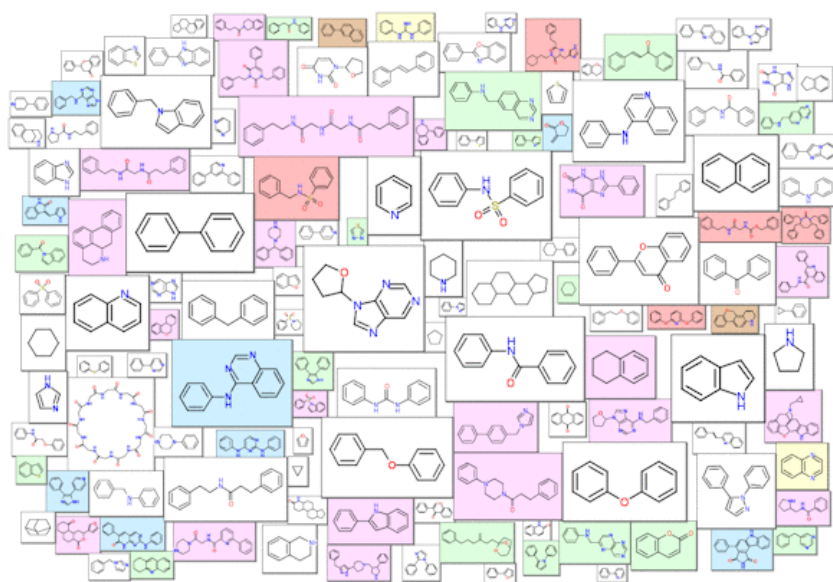
Combinatorial complexity at multiple scales

- Diverse constituents
- Arrangement of constituents
- Disordered

AI/ML models need readable data

How do we **represent** the materials to capture critical information?

Molecular mixture electrolytes

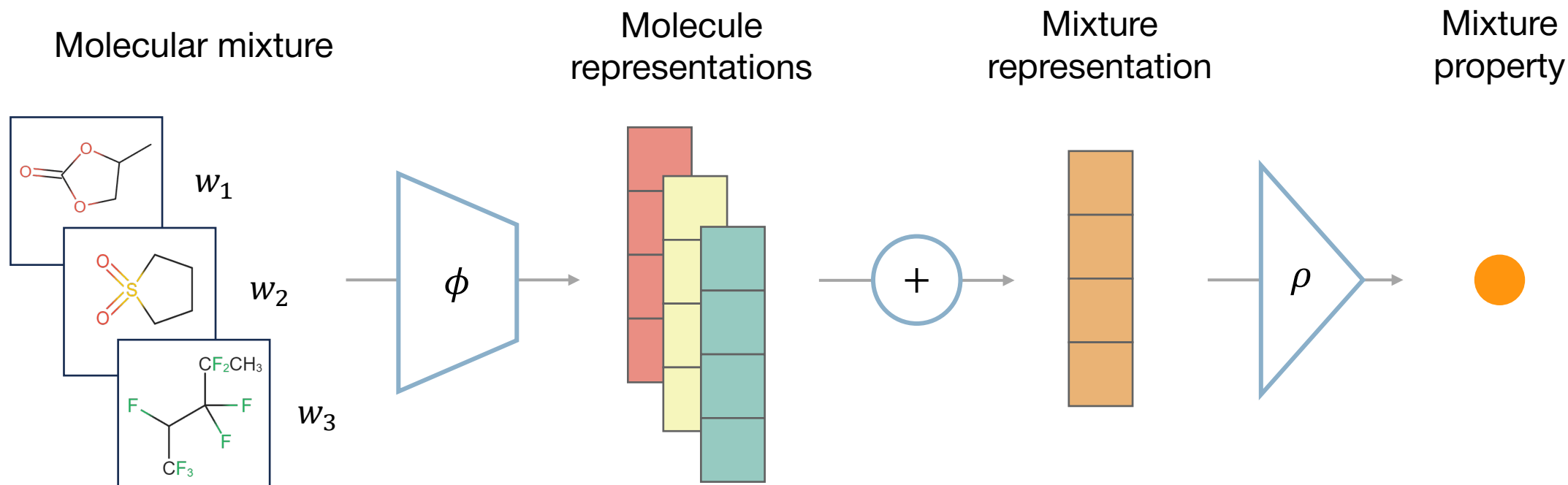


~160 molecules, ~40 salts

- $C(200,4) > 10^7$ compositions
- Infinite if considering fractions

- Molecules have diverse chemistry and geometries
- Mixture = constituents + molecular interactions

Modeling mixture as graph set



Graph neural network (GNN)

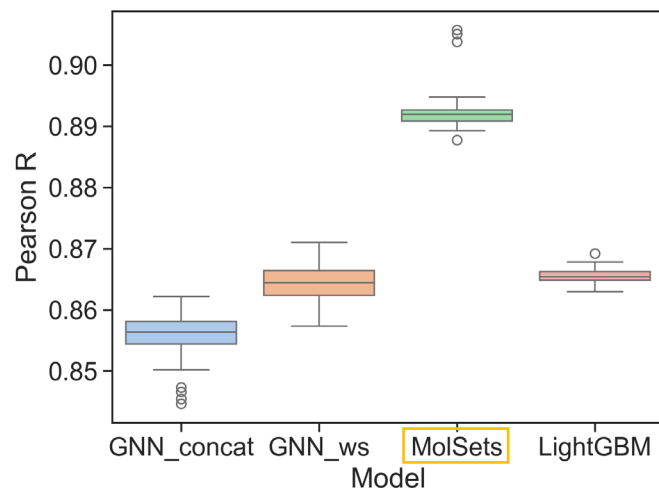
- Capture chemistry and geometry

Attention mechanism (as in GPT)

- Learns relative importances
- Permutation invariant

Benefits and applications

Superior accuracy in benchmark tests



Virtually screened >10,000 new candidate mixtures

Predicted room temperature electrical conductivity of molecular mixtures

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Author affiliations

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Citation

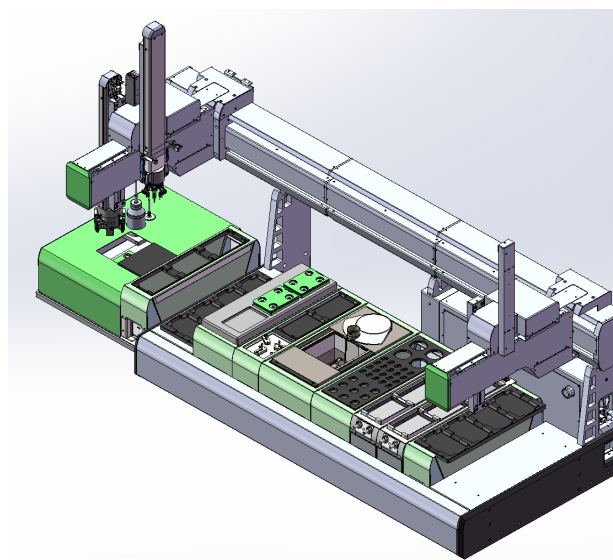
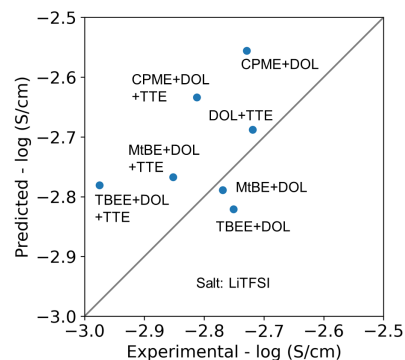
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Subject keywords

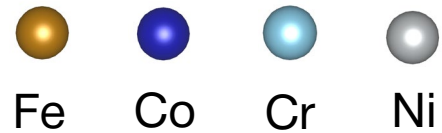
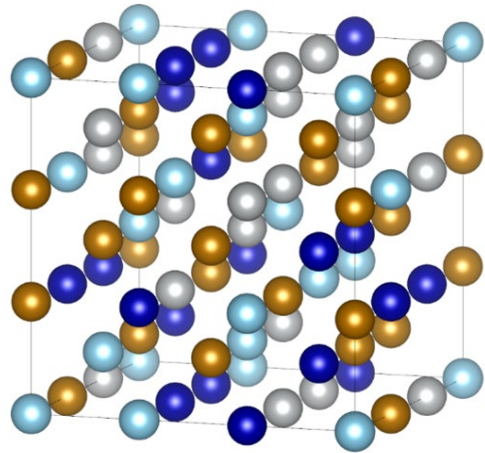
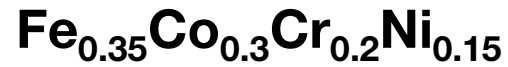
Exp. verified (Tianxing Lai @ UT Austin)



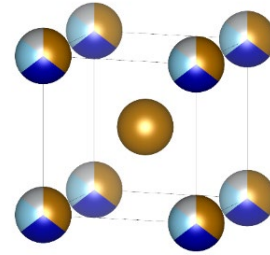
High-throughput,
autonomous experiments
(Jeffrey Lopez @ NU)

Extend to high-entropy alloys

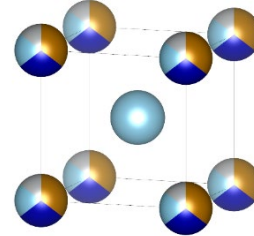
(Mentoring Ms. Ruishu Huang @ UW Madison)



HEA structure



+



+

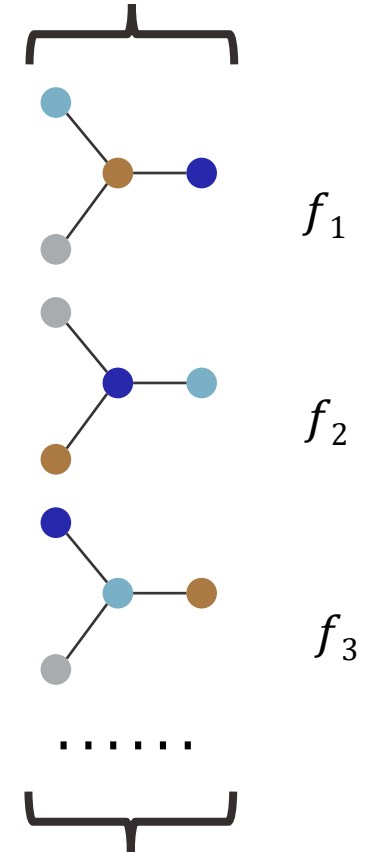
.....

$$f_1 = 0.35$$

(f : fraction)

$$f_2 = 0.2$$

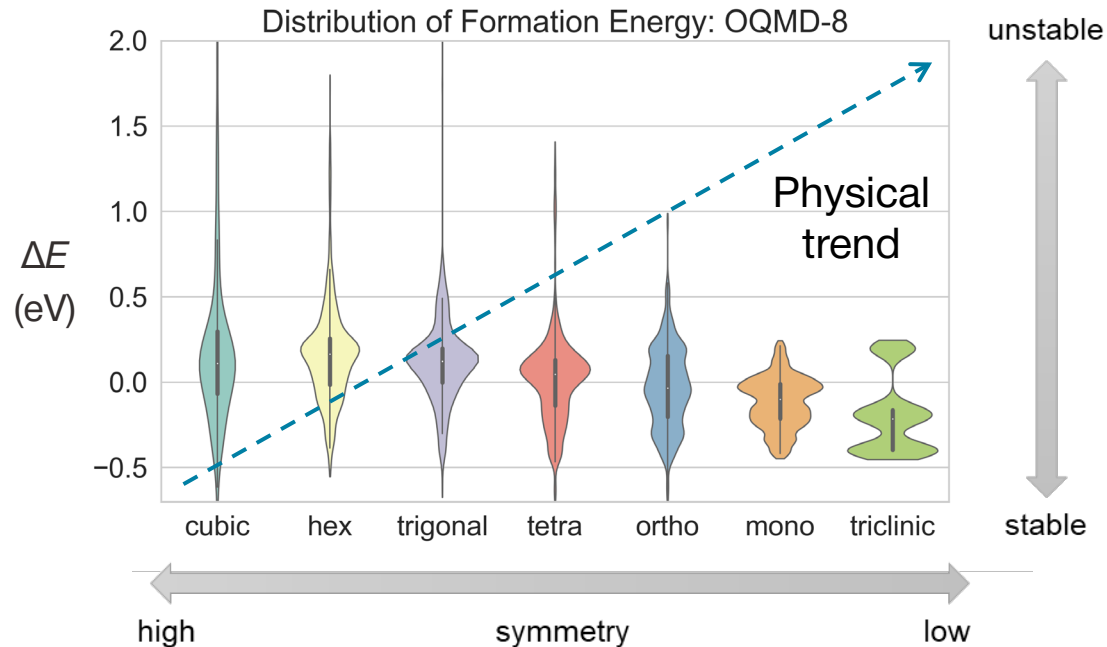
Collection of local environments (LEs)



LE graph set

Examine data behind ML models...

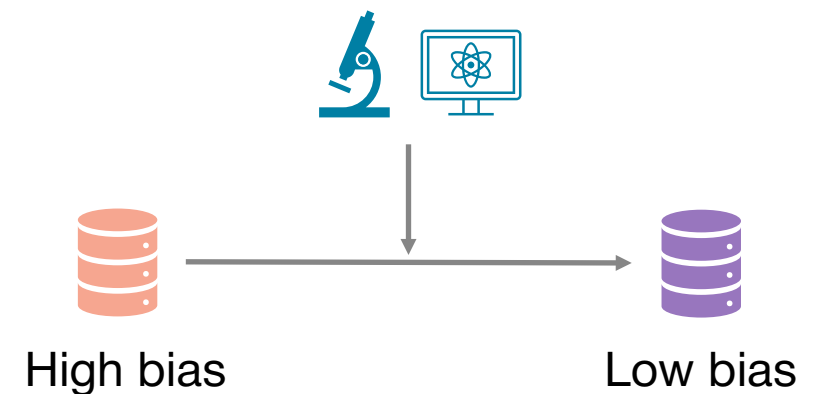
Data **bias** in materials data platforms, e.g., OQMD:



Source: unbalanced focus and knowledge
Biased data → problematic ML models.

Entropy-targeted active learning (ET-AL)

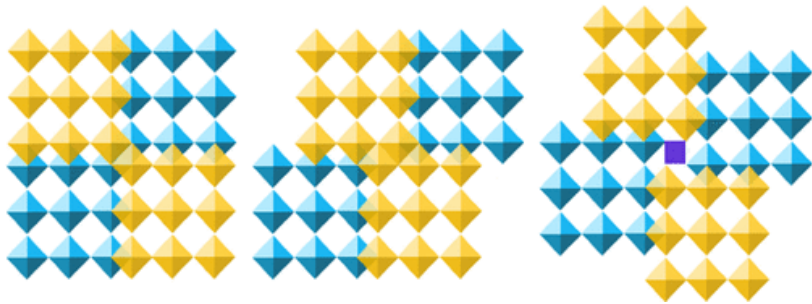
- A metric to quantify bias
- A framework to mitigate bias
 - Guide new data **acquisition**
- Construct high-quality database



Summary of contributions

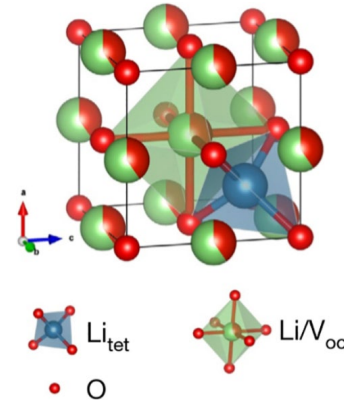
- ❖ Generic **representation** and ML model that address structural complexity
- ❖ Data **acquisition** for design in a broad composition–structure space

Combinatorial complexity formed by
diverse constituents $\times$ configurations

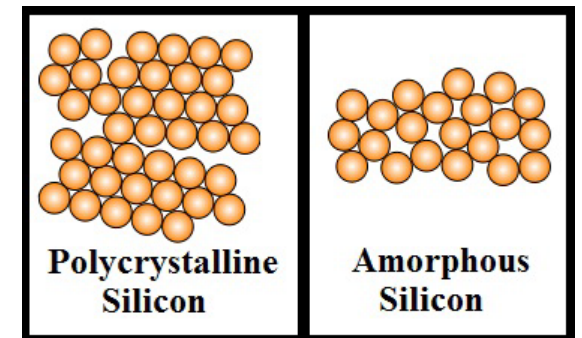


Digital superlattices

Disordered materials in energy and
electronic technologies

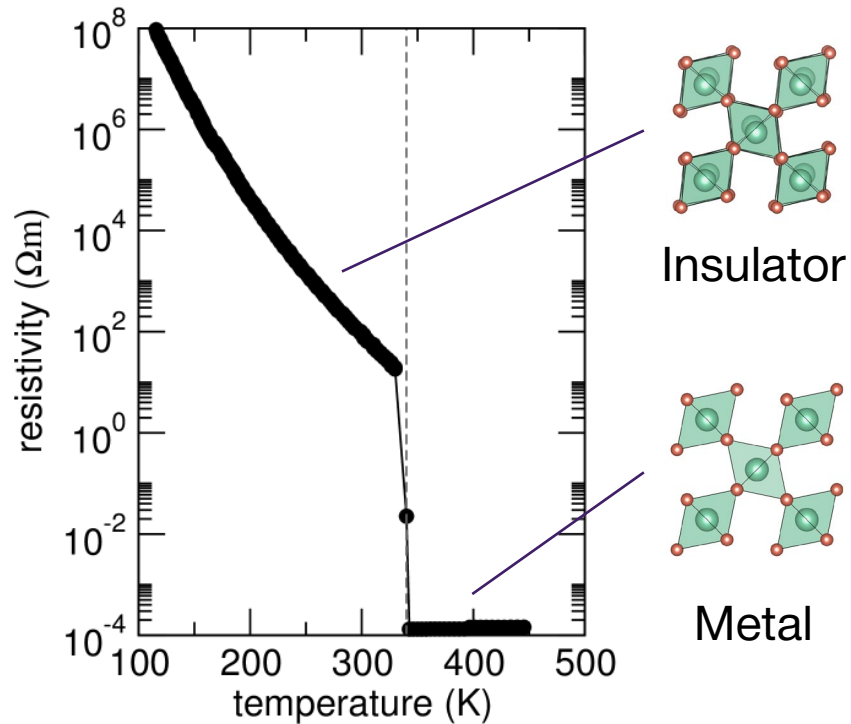


Disordered rock salt (DRX)
in Li batteries



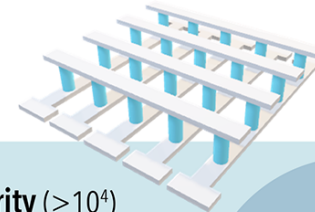
Amorphous semiconductor
in electronic devices

Designing combinatorial materials



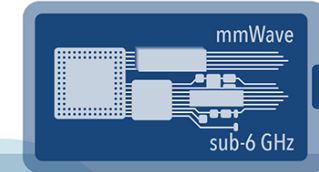
MIT materials are useful in microelectronic devices:

Cross-Point Selectors



High Non-linearity ($>10^4$)
 High I_{ON} ($>100\mu\text{A}$)
 Memory Specific V_T ($<1\text{V}$)
 Bipolar Symmetric Operation
 Cycling V_{TH} Drift ($<10\text{mV/dec}$)

Fast Switching Speed ($<10\text{ns}$)
 Fast Turn-ON ($<10\text{ns}$)
 High Endurance ($>10^{15}$)
 85°C Operation



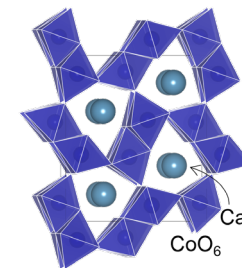
RF Switches & Limiters

High Linearity ($>30\text{dBm}$)
 Low Insertion Loss ($<1\text{dBm}$)
 High Isolation ($>30\text{dB}$)
 Tunable Threshold
 Power Handling ($>20\text{dBm}$)
 $>125^\circ\text{C}$ Operation

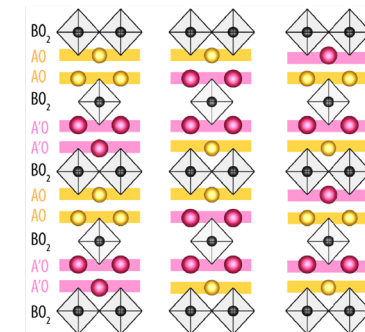
Metal-insulating transition (MIT) materials

- Tradeoff btw. resistivity change $\Delta\rho/\rho$ and temperature T_{MIT}

Single Phase Compounds



Digital Superlattices



Acknowledgments



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IDEAL Lab



Dr. James Rondinelli's
MTD Group



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@ NU



Daniel Apley
@ NU



Vinayak Dravid
@ NU



Elsa Olivetti
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A. Manthiram
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Divine Kumah
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@ IUB

