

# Yi Yang

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## EDUCATION

### Carnegie Mellon University

*Doctor of Philosophy (Ph.D.) in Materials Science and Engineering*

Pittsburgh, PA

*Jan. 2023 – Dec. 2026 (expected)*

### Carnegie Mellon University

*Master of Science in Computational Materials Science and Engineering*

Pittsburgh, PA

*Aug. 2021 – Dec. 2022*

### China University of Geosciences, Wuhan

*Bachelor of Science in Materials Science and Engineering (experimental class)*

Wuhan, China

*Sept. 2017 – July 2021*

## PUBLICATIONS

- [1] Gharakhanyan, V.\*, **Yang, Y.\***, et al. FastCSP: Accelerated Molecular Crystal Structure Prediction with Universal Model for Atoms. *arXiv:2508.02641* (2025). [Under review, *Nature Machine Intelligence*; \* Equal contribution.]
- [2] Gharakhanyan, V., **Yang, Y.**, et al. Open Molecular Crystals 2025 (OMC25) Dataset and Models. *Scientific Data* (2026)
- [3] **Yang, Y.**, et al. Genarris 3.0: Generating Close-Packed Molecular Crystal Structures with Rigid Press. *Journal of Chemical Theory and Computation* **21**, 11318–11332 (2025).
- [4] Nayal, K. S., **Yang, Y.**, et al. Efficient Molecular Crystal Structure Prediction and Stability Assessment with AIMNet2 Neural Network Potentials. *Crystal Growth & Design* (2025).
- [5] Hunnisett, L. M. et al. The Seventh Blind Test of Crystal Structure Prediction: Structure Ranking Methods. *Acta Cryst. B* **80**, 548–574 (2024).
- [6] Tom, R., **Yang, Y.**, et al. Inverse Design of Tetracene Polymorphs with Enhanced Singlet Fission Performance by Property-Based Genetic Algorithm Optimization. *Chemistry of Materials* **35**, 1373–1386 (2023).

## RESEARCH EXPERIENCE

### Research Assistant, Artificial Intelligence for Crystal Structure Prediction

*Carnegie Mellon University, Advisor: Noa Marom*

Pittsburgh, PA

*Sept. 2021 – Present*

### Accelerated Molecular Crystal Structure Prediction with Machine Learning Atomic Potential

- Constructed the **Open Molecular Crystals 2025 (OMC25)** dataset, over **27 million** molecular crystal structures across 12 elements, supporting large-scale training and evaluation of **machine learning interatomic potentials (MLIPs)**.
- Trained and benchmarked state-of-the-art MLIPs (**Universal Models for Atoms (UMA)**), achieving **DFT-level accuracy** and demonstrating superior generalization across targeted molecular crystal tasks.
- Co-developed **FastCSP**, an open-source **Python** crystal structure prediction (CSP) workflow integrating **Genarris 3.0** with **UMA** for high-throughput molecular crystal structure **generation** and **ranking**.
- Collaborated with the **FAIR Chemistry** team to publish and validate **MLIPs transferability** and **robustness** in large-scale FastCSP workflows on **drug-like** and other organic molecular crystals.

### Genarris 3.0: Random Molecular Crystal Structure Generator | (<https://github.com/Yi5817/Genarris>)

- Developed **Genarris 3.0**, an open-source **Python/C** package with **MPI-based parallelism** for large-scale molecular crystal structure generation on **high-performance computing (HPC)** clusters.
- Designed and implemented **Rigid Press**, a novel **geometry optimization algorithm** employing a **regularized hard-sphere potential** to compress molecular crystal unit cells without explicit energy evaluations.
- Integrated **GPU-accelerated MLIPs** and **clustering** for efficient down-selection in the CSP workflow, accelerating exploration of molecular crystal energy landscapes.
- Validated **robustness** of CSP workflow in Genarris 3.0 across multiple **drug-like** molecular crystals and **energetic materials**, supporting future applications in **AI-driven materials discovery**.

### The 7<sup>th</sup> Crystal Structure Prediction Blind Test, Cambridge Crystallographic Data Center

- Pioneered early application of MLIPs for CSP, achieving near-DFT accuracy in **ranking phase** with significantly reduced computational cost.
- Developed **system-specific AIMNet2 MLIPs** trained on DFT datasets of **molecular clusters (*n*-mers)** sampled from **MD** simulations, using **active learning** to improve accuracy and outperforming other ML-based CSP methods.

## PROJECTS

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- Crystal Structure Prediction with Diffusion Model** | 10-708, CMU *Sept. 2024 - Dec. 2024*
- Explored **autoregressive diffusion model** to enhance **conformer generation** accuracy for complex molecular systems.
- AI Ramanujan: Discovery of Formula Equivalence** | 11-785, CMU *Oct. 2022 - Dec. 2022*
- Trained **Transformer** encoders to identify equivalent mathematical formulas.
  - Designed a **translator** to convert mathematical formulas from Maple to Mathematica format.
  - Enhanced the encoders with **cross-attention** to capture correlations between embeddings.
  - Employed **contrastive learning** for model training to improve performance.
- JPX Tokyo Stock Exchange Prediction** *May 2022 - July 2022*
- Engineered key financial features from JPX data for trend forecasting.
  - Developed a **Long short-term memory (LSTM)** model with PyTorch for predicting JPX stock trends, ranking top **13%** on Kaggle.

## HONORS & AWARDS

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- Presidential Fellowship, College of Engineering, CMU** *Feb. 2026*
- Machine Learning in Chemical and Materials Sciences (MLCM) Rising Star Award, LANL** *May 2025*
- Frontera Computational Science Fellowship, TACC** *Apr. 2025*
- Award for Research Excellence in the Master Program, CMU** *May 2023*
- ATK-Nick G. Vlahakis Graduate Fellowship, CMU** *Feb. 2023*
- Outstanding Undergraduate Award, CUG** *June 2021*

## TECHNICAL SKILLS

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**Programming Languages:** Python, C, R

**Frameworks & Libraries:** PyTorch, TensorFlow, ASE, Pymatgen, RDKit, OpenBabel, PySpark

**Systems & Tools:** Linux, High-Performance Computing (HPC), MPI, AWS

**High Performance Computing:** Bridges2 (PSC), Frontera (TACC), Vista (TACC), Perlmutter (NERSC)

**Scientific Software:** FHI-aims, VASP, Orca

**Machine Learning Models:** Transformer, Graph Neural Network (GNN), LSTM

**Domains:** Scientific Computing, Computational Materials, Drug Discovery

**Languages:** English (Fluent), Chinese (Native Speaker)