

Xingyue Liu

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Profile

Dalian University of Technology

Dalian, China

Artificial Intelligence, School of Future Technology

2023-2027 Undergraduate

Academic Score: 87.0/100

Language Test: CET-4 618, CET-6 591

Publications

1. **Xingyue Liu**[†], Zijie. Xing[†], Runze. Wang, Luoming. Hu, and Yanming. Shen*, “Language-guided expertise evolution for protein optimization,” *ICLR 2026 Workshop RSI. (Accepted)* [\[Link\]](#)
2. R. Wang, Y. Ma, **Xingyue Liu**, Z. Xing, and Y. Shen*, “Motif-driven molecular graph representation learning,” *Expert Systems with Applications*, vol. 269, p. 126484, 2025. (Published) [\[Link\]](#)
3. Runze Wang, Zijie Xing, **Xingyue Liu**, Mingqi Yang, Che He, Yanming Shen*, “From Graphs to Tokens: Substructure-Aware Molecular Representation for Large Language Models,” *Information Processing and Management. (Published)* [\[Link\]](#)
4. Zijie Xing[†], **Xingyue Liu**[†], Runze Wang, Luoming Hu, Yanming Shen*, “KnowEvo: Knowledge Evolution for Protein Optimization,” *NeurIPS 2026. (Under Review)* [\[PDF\]](#)

Research Experience

04/2025 05/2026	Language-Guided Expertise Evolution for Protein Optimization Advisor: Prof. Yanming Shen Overview: Developed a self-evolving agent system for protein optimization that improves accuracy through iterative “explore–verify–refine” cycles, maintaining a population of code blocks as a knowledge base to accumulate executable prior knowledge. Core Contributions: (1) Code as Knowledge: Structured code populations serve as the knowledge base; each code block integrates a gating mechanism to match and optimize target protein sub-distributions. (2) Self-Evolving System: Built a high-concurrency self-evolving agent system from scratch, decoupling forward exploration from knowledge updates to improve efficiency, with multi-server architecture for better resource utilization. (3) Search Algorithm: Thompson sampling during training and LCB scoring at inference to balance exploration–exploitation in selecting the final knowledge base. Outcome: An early version won the Best Innovation Award at the Bohr+SciMaster AI4S Agent Development Competition (only 2 teams awarded) and was shortlisted for the Qizhi Startup Camp interview (top 10%). Published as co-first author at ICLR 2026 RSI Workshop [1] currently under review at NeurIPS 2026 [4] .
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10/2024 12/2024	Molecular Representation Learning Advisor: Prof. Yanming Shen Overview: Integrated high-level molecular motif information into both graph neural networks (GNNs) and large language models (LLMs) to enhance molecular representation learning, overcoming the expressive limits of 1-WL testing and 1D SMILES representations. Core Contributions: (1) Uni-Motif Framework (GNNs): Designed and evaluated three plug-and-play fusion mechanisms for incorporating molecular substructure information; conducted ablation studies on different motif decomposition strategies. (2) S^2 Token Strategy (LLMs): Proposed and systematically evaluated a novel substructure-aware tokenization method; reproduced multiple state-of-the-art baselines to validate the advantage of motif-augmented approaches. Outcome: Published as co-author in two peer-reviewed journals: <i>Expert Systems with Applications</i> [2] and <i>Information Processing and Management</i> [3].
12/2023 Present	Prediction of Protein Crystallization Conditions Advisor: Prof. Song Xue Overview: Predicted protein crystallization conditions by modeling the complex relationships among protein sequences, 3D structures, and chemical environments. Core Contributions: • Dataset Construction: Led the full project pipeline to build a crystallization condition database (~7k proteins), extracting, filtering, and integrating crystallization records from the PDB. • pLM Fine-tuning: Fine-tuned SaProt (a pretrained structure-aware protein language model) to capture latent relationships among protein sequence, structure, and crystallization conditions. • Crystallization Condition Similarity Analysis: Clustered proteins by structural similarity using Foldseek, analyzing combinatorial effects of solution components and the distribution of crystallization conditions under similar structures. Outcome: Experimental results indicate that pLMs struggle to effectively represent protein crystallization conditions, motivating the incorporation of hydration information into existing protein embedding models. Work currently in progress.

Awards & Honors

1. AY 2023–2024, AY 2024–2025 **Academic Excellence Scholarship**
2. AY 2024–2025 **National Scholarship**

Research Interests

- **Auto Scientific Research:** Improve the long-horizon scientific research ability of AI, focusing on iterating ideas with experimental feedback and proposing valuable ideas.

Skills

- **AI agent:** Self-evolving agents, agent harness design
- **AI for Science:** Protein representation learning, molecular representation learning, protein optimization