
JEFF GUO

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I am a final year PhD student at EPFL (**September 2025**) advised by Professor Philippe Schwaller, developing **molecular generative models** for **accelerated drug discovery**. Previously, I was an intern at [Microsoft Research AI4Science](#) with Jose Garrido Torres, Elise van der Pol, and Marwin Segler. Before that, I was a **cheminformatics researcher** in the [Molecular AI](#) department at AstraZeneca where I **developed tools for molecular design** and contributed to the **identification of experimentally validated candidate drug molecules**. My undergraduate degree was in chemistry, inspiring my research to **harness domain knowledge to augment machine learning methods** for chemical discovery.

RESEARCH EXPERIENCE

PhD Researcher, EPFL

Sep. 2022 –

Developing **sample-efficient molecular generative models** for drug and catalyst design with **synthesizability constraints**. Prospective experimental validation of generated molecules for binding. Applying Bayesian optimization for chemical reaction optimization with on-going collaboration with experimental chemists.

Research Intern, Microsoft AI4Science

Oct. 2023 – Jan. 2024

Machine learning for chemistry. Supervised by Jose Garrido Torres, Elise van der Pol, and Marwin Segler.

Graduate Scientist Researcher, Molecular AI AstraZeneca

Sep. 2020 – July 2022

Expanded the capabilities of the generative reinforcement learning molecular design tool, 'REINVENT', to accelerate early-phase drug discovery. Contributed to the identification of candidate drug molecules by coupling 'REINVENT' with structure-based molecular docking, ligand-based pharmacophore matching, and fragment linking functionalities. Devised and co-advised a research project to integrate active learning with 'REINVENT' to improve sample efficiency.

Research Assistant, Imperial College London

Nov. 2019 – Aug. 2020

Led small molecular inhibitor design for an oncology target by applying fragment-based strategies with molecular docking. Designed and synthesized candidate molecules and demonstrated binding via X-ray crystallography and NMR. Set up a virtual screening 'Glide' docking pipeline on an AWS cloud computing server, increasing throughput from the order of 10^4 to 10^7 molecules.

Wet-lab Research, McGill University and Alnylam Pharmaceuticals

May 2017 – Sep. 2019

Three research internships in organic synthesis for green and nucleic acid chemistry.

SELECTED OPEN-SOURCE PROJECTS

'Saturn' – [GitHub](#), [Paper](#)

Sample-efficient generative molecular design frame-work. Includes '[TANGO](#)' (Constrained synthesizability), '[General Synthesizability](#)' and, '[Steerable and Granular Synthesizability Control](#)' implementations.

'Augmented Memory' – [GitHub](#)

Fork of 'REINVENT' 3.2 with '[Augmented Memory](#)' and '[Beam Enumeration](#)' implementations for sample efficiency and explainability in generative molecular design.

'REINVENT' Version 3.2 – **Repository Contributor**, [GitHub](#)

Contributed to the expansion of 'REINVENT' capabilities with [Curriculum Learning](#) and [Linker Generation](#). Wrote tutorial notebooks ([GitHub](#)).

EDUCATION

PhD Chemistry and Chemical Engineering, EPFL

Sep. 2022 –

NSERC Post-Graduate Doctoral Scholarship – \$120,000

MRes Drug Discovery and Development, Imperial College London

Oct. 2019 – Sep. 2020

BSc Honours Chemistry, McGill University

2015 – 2019

SKILLS

Molecular design: Generative models with transfer learning/fine-tuning and goal-directed generation (structure-based and ligand-based). Virtual screening and with active learning for large libraries.

Molecular docking: Glide, gnina, AutoDock Vina, QuickVina2-GPU, MOE, GOLD, OpenEye

Ligand-based tools: ROCS (OpenEye), ESP-sim

Chemical reactions: Bayesian optimization, hands-on wet-lab experience (synthesis and characterization)

Quantum chemistry: xTB (semi-empirical), DFT (ORCA)

Chemical visualization: ChemDraw, PyMOL

Programming: Python, PyTorch, Linux, HPC, scikit-learn, RDKit

SUPERVISION

Victoire Lang '25 BSc, EPFL (Internship – NeurIPS 2024 AI4Mat)

July 2024 – Aug. 2024

Rémi Schlama '24 MSc, EPFL (Internship)

Mar. 2024 – May 2024

Sacha Raffaud '23 MSc, ICL (MSc [Thesis](#) – NeurIPS 2023 Diffusion Workshop)

June 2023 – Sep. 2023

Christian Josefson & Clara Nyman '22 MSc, Chalmers (MSc [Thesis](#))

Jan. 2022 – May 2022

SELECTED PUBLICATIONS (See [Google Scholar](#) for all publications)

§ denotes equal author contribution

- [1] J. Guo[§], V. Sabanza[§], Z. Jončev, J. Luterbacher, P. Schwaller “Generative Molecular Design with Steerable and Granular Synthesizability Control”, *arXiv*, 2025. [Link](#)
- [2] H. Wang, J. Guo, L. Kong, R. Ramprasad, P. Schwaller, Y. Du, C. Zhang “LLM-Augmented Chemical Synthesis and Design Decision Programs”, *ICML* 2025. [Link](#)
- [3] J. Guo, P. Schwaller “It Takes Two to Tango: Directly Optimizing for Constrained Synthesizability in Generative Molecular Design”, *Nat. Comput. Sci (accepted)*, 2026. [Link](#)
- [4] J. Guo, P. Schwaller “Directly Optimizing for Synthesizability in Generative Molecular Design using Retrosynthesis Models”, *Chem. Sci.*, 2025. [Link](#)
- [5] **Review Paper:** Y. Du[§], A. R. Jamasb[§], J. Guo[§], T. Fu, C. Harris, Y. Wang, C. Duan, P. Liò, P. Schwaller, T. L. Blundell “Machine Learning-Aided Generative Molecular Design”, *Nat Mach Intell*, 2024. [Link](#)
- [6] J. Guo, P. Schwaller “Saturn: Sample-efficient Generative Molecular Design using Memory Manipulation”, *Nat Mach Intell (accepted)*. [Link](#)
- [7] J. Guo, P. Schwaller “Beam Enumeration: Probabilistic Explainability For Sample Efficient Self-conditioned Molecular Design”, *ICLR* 2024. [Link](#)
- [8] J. Guo, P. Schwaller “Augmented Memory: Sample-Efficient Generative Molecular Design with Reinforcement Learning”, *JACS Au*, 2024. [Link](#)
- [9] M. Dodds, J. Guo, T. Löhr, A. Tibo, O. Engkvist, J. P. Janet “Sample Efficient Reinforcement Learning with Active Learning for Molecular Design”, *Chem. Sci.*, 2024. [Link](#)
- [10] J. Guo[§], F. Knuth[§], C. Margreitter, J. P. Janet, K. Papadopoulos, O. Engkvist, A. Patronov “Link-INVENT: Generative Linker Design with Reinforcement Learning”, *Digital Discovery*, 2023. [Link](#)
- [11] J. Guo[§], V. Fialková[§], J. D. Arango, C. Margreitter, J. P. Janet, K. Papadopoulos, O. Engkvist, A. Patronov “Improving De Novo Molecular Design with Curriculum Learning”, *Nat Mach Intell*, 2022. [Link](#)
- [12] J. Guo, J. P. Janet, M. R. Bauer, E. Nittinger, K. A. Giblin, K. Papadopoulos, A. Voronov, A. Patronov, O. Engkvist, C. Margreitter “DockStream: A Docking Wrapper to Enhance De Novo Molecular Design”, *J Cheminform*, 2021. [Link](#)