

# Wu, Renjie (Roger)

Boston, MA | [renjiewu@hsph.harvard.edu](mailto:renjiewu@hsph.harvard.edu) | +1 (857) 209-1097 | [linkedin.com/in/renjiewu](https://www.linkedin.com/in/renjiewu) | [ren-jie-wu.github.io](https://ren-jie-wu.github.io)

## ABOUT ME

Computational Biologist with a robust background in machine learning and biology. Strong mathematical basis, 4+ years of Python development experience and 2+ years of hands-on experience developing and applying ML models for applications. Skilled at working in collaborative, interdisciplinary environments. Seeking to leverage domain knowledge, research expertise and problem-solving skills to make real-world impacts in healthcare innovation with AI.

## EDUCATION

### Harvard University, School of Public Health

Boston, MA

M.S. in Computational Biology

08/2024 – 05/2026 (Expected)

- **GPA:** 3.8/4.0; **Courses:** Data Science, Computational Biology, High Performance Computing, Software Systems
- **Awards:** Harvard Chan Fellowship (2024)

### Peking University, School of Life Sciences

Beijing, China

B.Eng. in Bioinformatics

09/2020 – 07/2024

- Major **GPA:** 3.8/4.0; **Courses:** Data Structure & Algorithms, Machine Learning, Statistical Inference, Stochastic Process
- **Awards:** Award for Contribution in Student Organizations (2022), Qin Wanshun – Jin Yunhui Scholarship (2021)

## SKILLS

- **Domains:** Data Science, Machine Learning, Software Engineering, Statistical Analysis, Computational Genomics
- **Programming Languages:** Proficient in Python, R | Experienced in Bash, C/C++, HTML, JavaScript, SQL
- **Libraries/Tools:** PyTorch, Scikit-learn, Scipy, Scanpy, Flask, React, Git, Docker, Linux, AWS

## EXPERIENCES

### Protein–Ligand Interaction Network Classification Model Optimization

Boston, MA

Research Assistant in Dr. Giulia Menichetti's Lab, Harvard Medical School

03/2025 – 05/2025

- Collected, cleaned, and processed protein-ligand interaction datasets, employing network-based negative sampling methods to mitigate label distribution bias.
- Encoded proteins and ligands with pre-trained large language models, yielding significant improvements in the interaction classifier's transductive and inductive testing performance (AUROC = 0.920; AUPRC = 0.938).

### Web App Engineering of LLM-Powered Knowledge Management System

Boston, MA

Course Project in Architecting & Engineering Software System, MIT

11/2024 – 12/2024

- Developed independently a full-stack web application for end-to-end deployment using Flask (backend) and React (frontend) for a knowledge management system with a user-friendly interface.
- Integrated Large Language Model (LLM) API and Graph RAG techniques to enable effective information retrieval, enhancing the visualization of knowledge and theoretically improving retrieval accuracy by up to 3 times compared with the non-graph-based approaches.

### Single-Cell Deep Learning Model for Predicting Gene Expression Dynamics

Beijing, China

Undergraduate Researcher in Dr. Jinzhao Wang's Lab, Peking University

12/2023 – 05/2024

- Developed a machine learning model to predict gene expression changes over time using single-cell RNA velocity data.
- Built scalable data preprocessing pipelines for feature transformation, selection, and dimensionality reduction.
- Designed and implemented a hybrid deep learning architecture combining Graph Neural Networks and Transformer in Pytorch framework, optimized by regularization, hyper-parameter tuning, and ablation studies, achieving a 35% prediction accuracy improvement compared with the linear model.
- Applied the trained model to infer gene regulatory networks, reaching an AUC of 0.78, surpassing prior studies

### Multi-omics Deep Learning Model for Predicting Cancer Subtypes

Beijing, China

Research Assistant in Dr. Jinzhao Wang's Lab, Peking University

03/2023 – 06/2023

- Collaborated with computational specialists, contributing to a [publication](#) in *Briefings in Bioinformatics* as a co-author.
- Integrated biological insights to guide both architecture design and biological validation of a deep learning model framework for predicting cancer subtypes and markers from multi-model genomic datasets (~9,000 cells).