Mahdi (Sum) Qezlou, Ph.D.

Email: sumgezlou@gmail.com | (951) 565-6687 | linkedin.com/in/gezlou | gezlou.github.io

Innovative AI Scientist with a Ph.D. in Physics focusing on modeling complex systems with a strong foundation in statistics, math, and programming. Over 7 years of experience leading end-to-end multi-modal (time-series, images and 3D assets) machine learning projects from pre-training, fine tuning and Reinforcement Learning (RL). Proficient in Python-based ML libraries and scalable training and inference using 100+ TB of real-world data on cloud (Kubernetes) and HPC (Slurm).

Highlighted projects:

More: https://gezlou.github.io/ml.html

Foundation Model Training: Signal Detection in Trillion-Element, Noisy Astrophysical Datasets

Repos: https://github.com/qezlou/BlindSight

- **Objective**: Accurately identify a **subtle signal** from **noise and systematics** in trillion-resolution-element **image** and **time series data**, when traditional clustering methods proved ineffective.
- Contribution: Developed a domain-specific foundation model (Variational AutoEncoder
 (VAE)-Transformer-based architecture) for observed noisy time-series and images to model the underlying
 distribution of real signal as opposed to false positives. Leveraged advanced self-supervised learning
 techniques—including CLIP, DINOv2, MAE, BEIT, iBOT, and MoCo v3—employing contrastive learning, masked
 modeling, and self-distillation for robust vision and multimodal representation learning. This probabilistic model
 for false positive removal enabled percent-level accuracy in downstream analysis integral to the company's
 mission.

GenAl for inference: Fast surrogate models with Gen Al

Repos: https://github.com/gezlou/Goku ELG and https://github.com/gezlou/lila

- **Objective**: Enabling **accurate inference** through **fast surrogate models** replacing computationally prohibitive Astrophysical simulations
- Contribution: Led the development of probabilistic machine learning surrogate models, including stochastic models (e.g., Gaussian Processes) and Deep Learning generative models (Normalizing Flow), to enable robust causal inference in high-dimensional parameter space and improved forecasting accuracy by 50%.

Scalable GPU coding: GPU accelerated cosmological simulations

Repos: https://github.com/MP-Gadget/shenqi

- Objective: Adapt a massively parallel MPI–OpenMP CPU-based gravity and fluid solver for NVIDIA's
 Grace-Hopper heterogeneous CPU–GPU architecture deployed on the VISTA supercomputing platform.
 tacc.utexas.edu/systems/vista/
- Contribution: Led key aspects of low-level code migration and performance optimization, for multi-node GPU acceleration through novel CUDA-based neighbor-tree construction and FFT techniques to achieve 5X computational speedups.

Technical Skills

- **Theoretical Knowledge:** Solid understanding of Linear Algebra, probabilistic modeling and physics simulations.
- ML Tools: Experienced in large-scale training (100+ TB) using PyTorch and TensorFlow, with expertise
 in domain adaptation, e.g. NVIDIA's NeMo framework, including LSTM and Transformer
 encoder-decoder architectures.
- Programming skills: Distributed CPU-GPU computing on Cloud (Kubernetes) and HPC (Slurm, CUDA-aware MPI), NVIDIA's SDK (NeMo, cuda, etc.), python, C/++, Git, Linux.
- Soft Skills: Extensive interdisciplinary applied machine learning research experience, with a proven ability to collaborate cross-functionally with management and business stakeholders to align technical solutions with strategic objectives.

Employment

Al Research Scientist | University of Texas, Austin | 2024 – Present

ML Research Scientist | Carnegie Institute for Science | 2020 – 2024

ML Research Scientist | University of California, Riverside | 2018 – 2024

Education: