Li Ju

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TL; DR

Final-year Ph.D. candidate in Scientific Computing with expertise in federated learning, uncertainty quantification, and multi-modal language models. Seeking to apply advanced machine learning research and development skills to solve complex challenges in an AI Researcher or Data Scientist role.

EDUCATION

Uppsala University Ph.D. in Scientific Computing	Uppsala, Sweden 09.2021 – 05.2026 (Expected)
Uppsala University M.Sc. in Computational Science	Uppsala, Sweden 09.2019 - 08.2021
University of Science and Technology of China M.Sc. in Chemometrics, with Honors B.Sc. in Chemistry	Hefei, China 09.2016 - 06.2019 09.2012 - 06.2016

Work Experience

Doctoral researcher

09.2021 - Present

Scientific Machine Learning Laboratory

- Uppsala, Sweden
- Designed and implemented novel algorithms for fair federated learning and uncertainty quantification in large-scale vision-language models, leading to multiple publications.
- Pioneered methods for modelling aleatoric and epistemic uncertainty in multi-modal AI using probabilistic and generative approaches (e.g., hyperspheric probabilistic modeling, Riemannian Flow Matching).
- Contributed to research on attack/defence mechanisms in distributed ML and on hierarchical information management systems.
- Instructed graduate courses in *Data Engineering* and *Cloud Computing* and supervised three M.Sc. projects.

Master thesis student

02.2021 - 08.2021

Integrative Scalable Computing Laboratory

 $Uppsala,\ Sweden$

• Designed and implemented a proactive Kubernetes autoscaler using predictive modeling, reducing average workload response time by 15% in heterogeneous edge computing environments.

Junior data scientist (Part-time)

06.2020 - 03.2021

Scaleout Systems AB

Uppsala, Sweden

- Integrated PyTorch support into FEDn, a production-grade open-source federated learning framework, enhancing its core functionality and user adoption.
- Developed proof-of-concept applications in computer vision to demonstrate framework capabilities to potential clients.

Skills & Tools

Modeling: Uncertainty quantification, generative AI, multi-modal language models, federated learning, probabilistic modeling, Bayesian inference, self-supervised learning.

Frameworks: PyTorch, JAX, pthread, OpenMP, MPI, CUDA, Apache Pulsar.

Cloud & DevOps: Docker, Apptainer, Kubernetes, OpenStack.

Languages: Python, C/C++, Haskell, Erlang, Lisp/Racket.

Accelerating fair federated learning

- Analyzed the suboptimal convergence rates of first-order methods in existing federated learning formulations. Reformulated fairness-aware FL as a dynamic multi-objective optimization problem to achieve provably faster convergence while upholding fairness constraints.
- Proposed AdaFedAdam, an algorithm with adaptive hyperparameter tuning and normalized updates, providing theoretical guarantees for accelerated convergence and reduced fairness bias.
- Empirically validated the method's Pareto optimality and robustness across standard FL benchmarks under various data and system heterogeneity scenarios.

Aleatoric uncertainty quantification for vision language models

- Investigated the limitations of post-hoc uncertainty quantification methods for pre-trained vision-language models.
- Proposed and empirically validated a hypothesis on the structural asymmetry of uncertainty between vision and language modalities.
- Developed a novel framework to model aleatoric uncertainty in unit hyperspherical space, systematically evaluated against other state-of-the-art methods.
- Demonstrated enhanced model reliability on downstream tasks, including zero-shot classification and image retrieval, by integrating quantified uncertainty.

Epistemic uncertainty quantification for vision language models

- Investigated the limitations of existing uncertainty quantification methods for pre-trained vision-language models and identified the need for improved epistemic uncertainty modeling.
- Proposed a generative framework using Riemannian flow matching to explicitly model epistemic uncertainty in VLMs.
- Validated the approach through extensive experiments on standard VLM benchmark datasets and demonstrated significant improvements in model robustness and reliability, particularly for out-of-distribution detection.

SELECTED PUBLICATIONS

- Ju, L., Andersson, M., Fredriksson, S., Glöckner, E., Hellander, A., Vats, E., & Singh, P. (2025). Exploiting the Asymmetric Uncertainty Structure of Pre-trained VLMs on the Unit Hypersphere. Advances in Neural Information Processing Systems 2025.
- Ju, L., Zhang, T., Toor, S., & Hellander, A. (2024). Accelerating fair federated learning: Adaptive federated adam. IEEE Transactions on Machine Learning in Communications and Networking.
- Ju, L., Hellander, A., & Spjuth, O. (2024). Federated learning for predicting compound mechanism of action based on image-data from cell painting. Artificial Intelligence in the Life Sciences.
- Ju, L., Singh, P., & Toor, S. (2021). Proactive autoscaling for edge computing systems with kubernetes. Proceedings of the 14th IEEE/ACM International Conference on Utility and Cloud Computing Companion.
- Ju, L., Lyu, A., Hao, H., Shen, W., & Cui, H. (2019). Deep learning-assisted three-dimensional fluorescence difference spectroscopy for identification and semiquantification of illicit drugs in biofluids. Analytical chemistry.

AWARDS & LEADERSHIP

- 1st Place in Huawei Sweden Hackathon (2025): Developed a neural operator for SVD approximation, outperforming 78 teams from across Europe.
- 2nd Place in Huawei Sweden Hackathon (2024): Solved a wireless localization problem using channel charting, competing against over 30 European teams.
- **President** of Society for Industrial and Applied Mathematics (SIAM), Uppsala University Chapter (2024 Present). Organized seminars and workshops on numerical analysis and scientific computing.