

WENZHI FANG

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EDUCATION

Purdue University Elmore Family School of Electrical and Computer Engineering Ph. D. Candidate in Electrical and Computer Engineering	<i>West Lafayette, IN, US</i> <i>Aug. 2023 – 2027</i>
ShanghaiTech University School of Information and Science Technology M.S. in Communication and Information Systems	<i>Shanghai, China</i> <i>Sept. 2020 – Jul. 2023</i>
Shanghai University School of Communication and Information Engineering B.S. in Communication Engineering	<i>Shanghai, China</i> <i>Sept. 2016 – Jul. 2020</i>

MAJOR COURSES

- Machine Learning, Deep Learning, Reinforcement Learning
- Bayesian Data Analysis, Convex Optimization, Matrix Computation

TECHNICAL SKILLS

Languages	Python, C, C++, Git, Matlab, LaTeX
Tools	PyTorch, Tensorflow, vLLM, VERL

RESEARCH INTERESTS

- **Large Language Model.** Efficient Fine-Tuning of LLM & RL-based Post-training, LLM Agents
- **Machine Learning and Optimization.** Federated Learning & Distributed Optimization

WORKING EXPERIENCE

ION Lab	<i>Aug., 2023 - Present</i>
PhD Research Assistant	<i>Advisor: Prof. Christopher G. Brinton</i> <i>Purdue University</i>
• Collaborative Device-cloud LLMs Reasoning,	<i>May. 2025 – Sept. 2025</i>
We propose a unified reinforcement learning framework that empowers on-device LLMs to autonomously learn both problem-solving reasoning and routing policies, deciding when to process queries locally or to offload them to cloud LLMs. Our method removes the need for separate routing classifiers.	
• Multi-modal LLM Fine-tuning,	<i>May. 2025 – Sept. 2025</i>
we propose two-stage adaptive personalization (TAP), a federated learning framework that personalizes multi-task and multi-modal LLMs by integrating selective local adaptation with post-FL knowledge distillation to balance task-specific performance and generalization.	
• On-device LLM Fine-tuning,	<i>Oct. 2024 – Feb. 2025</i>
We propose sketching LoRA, a theoretically-grounded parameter efficient fine-tuning methodology that retains LoRA's flexibility while adapting to the communication and computational abilities of individual devices.	
• Distributed Optimization for Fog Learning,	<i>Jan. 2024 – June. 2024</i>
We proposed an algorithm to address multi-level data heterogeneity in hierarchical federated learning, a.k.a. fog learning, deriving strong theoretical results without relying on additional data heterogeneity assumptions. This work fills a critical gap in the existing federated learning literature.	

Machine Learning and Optimization Lab

Aug., 2022 - Feb. 2023

Summer Intern

Advisor: Prof. [Peter Richtarik](#)

KAUST

- **Primal-Dual Optimization,**

Aug., 2022 - Feb. 2023

We studied a primal-dual hybrid gradient descent algorithm and explored variance reduced technique for primal dual optimization.

Optimization for Communication Lab

Sept., 2020 - June. 2022

Master Research Assistant

Advisor: Prof. [Yong Zhou](#)

ShanghaiTech

- **Zeroth-order Optimization for Federated Learning,**

Jul. 2021 – Jan. 2022

We proposed a federated zeroth-order algorithm (FedZO) with a convergence guarantee. This algorithm makes the training process forward-only, eliminating the memory overhead of backward propagation, which has been widely applied in LLM related research, e.g., LLM attack, prompt tuning, and efficient model fine-tuning.

TEACHING EXPERIENCE

SI263: Distributed Optimization

Spring, 2022, ShanghaiTech

ACADEMIC SERVICE

Reviewer of top tier conferences including NeurIPS, ICML, ICLR, AISTAT, AAAI

Reviewer of top tier journals including IEEE/ACM Transactions On Networking; IEEE Transactions on Signal Processing (TSP); IEEE Transactions On Mobile Computing (TMC); Transactions on Machine Learning Research (TMLR)

SELECTED PUBLICATIONS

- **W. Fang**, D-J. Han, L. Yuan, and C. G. Brinton, “Collaborative Device-Cloud LLM Inference through Reinforcement Learning”, 2025, *Under Review* [[pdf](#)]
- **W. Fang**, D-J. Han, L. Yuan, S. Hosseinalipour, and C. G. Brinton, “Federated Sketching LoRA: On-Device Collaborative Fine-Tuning of Large Language Models”, 2025, *Under Review* [[pdf](#)]
- J. Lee, **W. Fang**, D-J. Han, S. Hosseinalipour, and C. G. Brinton, “TAP: Two-Stage Adaptive Personalization of Multi-task and Multi-Modal Foundation Models in Federated Learning”, 2025, *Under Review* [[pdf](#)]
- **W. Fang**, D-J. Han, and C. G. Brinton, “Federated Learning over Hierarchical Wireless Networks: Training Latency Minimization via Submodel Partitioning”, *IEEE/ACM Transactions On Networking (ToN)*, 2025 [[pdf](#)]
- Y. Zhou, **W. Fang**, Y. Shi, and K. B. Letaief “Federated Edge Learning: Algorithms, Architectures and Trustworthiness”, *Springer Nature*, 2025 [[pdf](#)]
- **W. Fang**, D-J. Han, E. Chen, S. Wang, and C. G. Brinton, “Hierarchical Federated Learning with Multi-Timescale Gradient Correction”, *Neural Information Processing Systems (NeurIPS)*, 2024. [[pdf](#)]
- D-J. Han, **W. Fang**, S. Hosseinalipour, and C. G. Brinton, “Orchestrating federated learning in space-air-ground integrated networks: Adaptive data offloading and seamless handover” *IEEE Journal on Selected Areas in Communications (JSAC)*, 2024 [[pdf](#)]
- **W. Fang**, Z. Yu, Y. Jiang, Y. Shi, C. Jones, and Y. Zhou, “Communication-Efficient Stochastic Zeroth-Order Optimization for Federated Learning”, *IEEE Transactions on Signal Processing (TSP)*, 2022 [[pdf](#)]