

# Tiansheng Huang

PhD student at Georgia Institute of Technology

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## Education

**Georgia Institute of Technology**, Atlanta, USA Aug 2022 – Present

- Third year PhD student, School of Computer Science
- Program Advisor: Prof. Ling Liu

**South China University of Technology**, Guangzhou, China Sept 2019 – June 2022

- M.S, School of Computer Science
- Program Advisor: Prof. Weiwei Lin
- Thesis: Application of Multi-arm Bandit Algorithms in Client Selection of Federated Learning

**South China University of Technology**, Guangzhou, China Sept 2015 – June 2019

- B.S, School of computer science
- GPA: 3.75 (rank top 10%)

## Research Interest

### Current interest

- My current research interest lies in security/privacy aspect of machine learning, distributed machine learning and parallel and distributed computing.

### Previous studied

- Multi-arm bandit
- Online learning
- Resource scheduling on cloud/edge computing

## Industrial Experience

**Dolby Advanced Technology Group**, Atlanta, USA May 2024 - August 2024

### *Research Intern*

- Refine service delivery pipeline for LLMs/VLMs against harmful fine-tuning.
- Program Advisor: Gautam Bhattacharya, Pratik Joshi, Josh Kimball

**JD explore academy**, Beijing, China March 2022 - June 2022

### *Research Intern*

- Develop Personalized FL algorithms with factorization and sparse compression.
- Program Advisor: Li Shen

**JD explore academy**, Beijing, China June, 2021 - Sept 2021

### *Research Intern*

- Develop high efficiency sparse training algorithms for personalized FL.
- Program Advisor: Li Shen

## Publications

### Peer-review Conference

[1] **T. Huang**, S. Hu, L. Liu, "Vaccine: Perturbation-aware Alignment for Large Language Model against Harmful Fine-tuning," **NeurIPS2024**

[2] **T. Huang**, S. Hu, F. Ilhan, S. Tekin, L. Liu, "Lazy Safety Alignment for Large Language Models against Harmful Fine-tuning," **NeurIPS2024**

- [3] S. Tekin, F. Ilhan, T. Huang, S. Hu, L. Liu, “LLM-TOPLA: Efficient LLM Ensemble by Maximising Diversity,” **EMNLP 2024 (Findings)**
- [4] K. Chow, S. Hu, **T. Huang**, L. Liu, “Personalized Privacy Protection Mask Against Unauthorized Facial Recognition”, **ECCV2024**.
- [5] K. Chow, S. Hu, **T. Huang**, F. Ilhan, W. Wei, L. Liu, “Diversity-driven Privacy Protection Masks Against Unauthorized Face Recognition”, **PET2024**.
- [6] F. Ilhan, G. Su, S. Tekin, **T. Huang**, S. Hu, L. Liu, “Resource-Efficient Transformer Pruning for Finetuning of Large Models”, **CVPR2024**.
- [7] S.Hu, **T. Huang**, KH. Chow, W. Wei, Y. Wu, L. Liu. “ZipZap: Efficient Training of Language Models for Ethereum Fraud Detection”, **WWW2024**.
- [8] F. Ilhan, KH. Chow, S. Hu, **T. Huang**, S. Tekin, W. Wei, Y. Wu, M. Lee, R.Kompella, H. Latapie, G. Liu, L. Liu, “Adaptive Deep Neural Network Inference Optimization with EENet,” **WACV2024**.
- [9] **T. Huang**, S. Hu, KH. Chow, F. Ilhan, S. Tekin, L. Liu, “Lockdown: Backdoor Defense for Federated Learning with Isolated Subspace Training,” **NeurIPS2023**.
- [10] Y. Sun, L. Shen, **T. Huang**, and D. Tao, “FedSpeed: Larger Local Interval, Less Communication Round, and Higher Generalization Accuracy,” **ICLR2023**.
- [11] F. Ilhan, SF Tekin, S Hu, **T. Huang**, KH Chow and L Liu, “Hierarchical Deep Neural Network Inference for Device-Edge-Cloud Systems[C]” **WWW2023**.
- [12] S. Hu, **T. Huang**, F. Ilhan, SF. Tekin, L. Liu, “Large Language Model-Powered Smart Contract Vulnerability Detection: New Perspectives” **IEEE TPS2023**.
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## Journal

- [1] **T. Huang**, L. Shen, Y. Sun, W. Lin, and D. Tao, “Fusion of Global and Local Knowledge for Personalized Federated Learning,” 2022, Transactions on Machine Learning Research (**TMLR**).
- [2] **T. Huang**, W. Lin, L. Shen, K. Li and A. Y. Zomaya, “Stochastic Client Selection for Federated Learning with Volatile Clients,” 2022, IEEE Internet of Things Journals (**IoT-J**).
- [3] **T. Huang**, W. Lin, X. Hong, X. Wang, Q. Wu, R. Li, CH. Hsu, AY. Zomaya, “Adaptive Processor Frequency Adjustment for Mobile Edge Computing with Intermittent Energy Supply”, 2021, IEEE Internet of Things Journals (**IoT-J**).
- [4] **T. Huang**, W. Lin, W. Wu, L. He, K. Li and AY. Zomaya, “An Efficiency-boosting Client Selection Scheme for Federated Learning with Fairness Guarantee,” 2020, IEEE Transactions on Parallel and Distributed Systems (**TPDS**).
- [5] **T. Huang**, W. Lin, C. Xiong, R. Pan and J. Huang, “An Ant Colony Optimization Based Multi-objective Service Replicas Placement Strategy for Fog Computing,” 2020, IEEE Transactions on Cybernetics (**TCYB**).
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## Under Submission

- [1] **T. Huang**, S. Hu, W. Wei, L. Liu, “Silencer: pruning-aware backdoor defense for decentralized federated learning,” Under Submission.
- [2] **T. Huang**, G. Bhattacharya, P. Joshi, J. Kimball, L. Liu, “Antidote: Post-fine-tuning Safety Alignment for Large Language Models against Harmful Fine-tuning,” Under Submission.
- [3] **T. Huang**, S. Hu, F. Ilhan, S. Tekin, L. Liu, “Booster: Tackling Harmful Fine-tuning for Large Language Models via Attenuating Harmful Perturbation,” Under Submission.
- [4] **T. Huang**, S. Hu, F. Ilhan, S. Tekin, L. Liu, “Harmful Fine-tuning Attacks and Defenses for Large Language Models: A Survey,” Under Submission.

## Projects

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### Area 1: Safety alignment for Large language models (Current focus)

#### 1.1 Alignment-Stage defense for LLMs’ harmful finetuning

-Uncover the reason of failure of safety alignment after fine-tuning an LLM on partially harmful data.

-Based on the reason of failure, which we name “harmful embedding drift”, we develop an alignment stage defense solution, which “vaccinate” the model to be immune of harmful finetuning.

### 1.2 Finetuning-Stage defense for LLMs’ harmful finetuning

- Develop a fine-tuning stage prototype solution for preserving safety alignment after harmful finetuning.
- Observed that *excess drift* towards the switching point might be the performance bottleneck for the prototype solution.
- Develop a refined solution by adding a proximal term to control the excess drift.

## **Area 2: Security aspect of Federated Learning (Previous study)**

### 2.1 Backdoor Defense for Federated Learning with isolated subspace training

- First to identify poison coupling effect in federated learning.
- Invent isolated subspace training technique to decouple and filter the poisoned parameters.
- Source code available at <https://github.com/LockdownAuthor/Lockdown>.

### 2.2 Pruning-aware Backdoor Defense for Decentralized Federated Learning

- Theoretically identify empirical Fisher information as a reliable indicator of poisoned parameters.
- Empirically study the Fisher-guided pruning technique to purify the poisoned model .
- Invent a defense to boost pruning-awareness in the training phase.

## **Area 3: Efficient Federated Learning/Personalized Federated Learning (Previous study)**

### 3.1 Efficient client selection in FL with multi-arm bandit

- Identify system heterogeneity/selection fairness/ cumulative participation as main factors for federated learning system performance.
- balance system heterogeneity/selection fairness/cumulative participation with UCB/stochastic multi-arm bandit algorithms.

### 3.2 Efficient PFL with low-rank+sparse

- Low-rank+sparse joint compression for personalized federated learning.
- Design a proximal algorithms.to solve the problem with theoretical guarantee.

### 3.3 Efficient PFL with dynamic sparse training

- Formulate the personalized models as a nested network in the global model.
- Propose a dynamic sparse training technique for training time acceleration in PFL.

## **Area 4: Resource scheduling in cloud/edge environment (Previous study)**

#### 4.1 Resource scheduling for renewable-energy supply edge devices

- Study computation offloading in a scenario that the edge devices are powered by renewable-energy supply.
- Formulate the problem as an event driven semi Markov decision process
- Solve the problem with a deep reinforcement learning technique.

#### 4.2 Service replicas placement in Fog computing

- Study replicas placement problem in Fog computing
- Formulate the problem as a Mixed Integer Linear Problem
- Solve the problem with an ant colony algorithm.

#### **Honor and Awards**

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- IEEE TPS 2023 student travel grant	2023
- National Scholarship (Top graduate scholarship in China)	2021
- National Scholarship	2020
- First-Class School Scholarship	2019

#### **Academy Service**

**Conference Reviewer:** NeurIPS (2023,2024), ICLR (2024,2025), ICML2024, AAAI2024

**Journal Reviewer:** TMC, TCOM, TP, TOIT, TMLR