

Xiaolan Gu (She/Her)

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EDUCATION

- **University of Arizona** *Tucson, AZ* 2018 - 2024
Ph.D., Electrical and Computer Engineering (GPA: 4.00/4.00)
- **Beihang University** *Beijing, China* 2015 - 2018
M.S., Automation Science and Electrical Engineering (GPA: 3.94/4.00)
- **Beihang University** *Beijing, China* 2011 - 2015
B.S., Mathematics and Systems Science (GPA: 3.58/4.00)

EXPERIENCE

- **Data Scientist, LinkedIn** *Sunnyvale, CA* July 2023 - Present
 - ◊ **Data Privacy Team: Differential Privacy, Federated Learning**
- **Applied Research Data Science Intern, LinkedIn** *Sunnyvale, CA* Summer 2022
 - ◊ **Generating Synthetic Tabular Data with Differential Privacy**
 - Motivation: answering a set of queries with differential privacy suffers from poor privacy-utility tradeoff (due to privacy budget splitting).
 - Project work: implemented the state-of-the-art mechanisms for synthetic data generation with differential privacy (DP), applied the developed software tool to LinkedIn's real datasets for privacy-preserving data analytics.
- **Research Intern, Security Lab, Baidu** *Sunnyvale, CA* Summer 2019
 - ◊ **Key-value Data Collection under Local Differential Privacy**
 - We developed a novel framework that utilizes correlated perturbations (between key and values) to enhance the privacy-utility tradeoff under local differential privacy (LDP).
 - We proposed an optimized privacy budget allocation approach with closed-form solutions to further improve the utility.
- **Graduate Research Assistant, University of Arizona** *Tucson, AZ* Fall 2018 - Spring 2023
Advisor: Dr. Ming Li
 - ◊ **Privacy-preserving and Robust Federated Learning**
 - Motivation: comparing with centralized machine learning, federated learning with differential privacy (DP) suffers from both poor privacy-utility tradeoff and vulnerability to model poisoning attacks.
 - We developed a hybrid solution (with both DP and crypto), which achieves better privacy-utility tradeoff and is resistant to client collusion. By leveraging Secure Multiparty Computation (MPC) techniques, the Gaussian noise added for privacy purpose is shown to provide robustness guarantee against model poisoning attacks.
 - ◊ **Input-Discriminative Local Differential Privacy**
 - We proposed a new privacy notion called Input-Discriminative LDP (ID-LDP), which is shown to be a fine-grained version of LDP and only needs less perturbation (and thus yields better utility) than LDP.
 - We developed two mechanisms for single-item input and item-set input on the application of frequency estimation, where both two mechanisms are shown to satisfy ID-LDP.
 - ◊ **On Hybrid Queries under Local Differential Privacy**
 - We proposed a novel and practical protocol for location-based applications to simultaneously enhance the utility for record-level queries and statistical/aggregated analysis.
 - We analyzed the theoretical Mean Squared Error of our protocol and showed the relationship to an existing protocol.

SKILLS

- Python (numpy, scipy, pandas, pytorch, scikit-learn), Matlab, Scala, C/C++, Git, SQL
- Optimization, Statistics, Algorithms, Machine Learning, Federated Learning
- Privacy-preserving techniques: Differential Privacy (DP) and Secure Multi-Party Computation (MPC)
- Robust machine learning (and federated learning) against adversarial examples and poisoning attacks

- Preprints

- ## PROFESSIONAL SERVICES

- ## AWARDS AND HONORS

- ## TEACHING

- Teaching Assistant, Computer Programming for Engineering Applications (C language) 2018 - 2019

RELEVANT COURSEWORK

- Machine Learning Theory; Online Learning and Multi-armed Bandits; Fundamentals of Data Science for Engineers.
- Nonlinear Optimization; Information Theory; Probability and Random Processes for Engineering; Fundamentals of Information and Network Security; Fundamentals of Computer Network.