

# Xiaolan Gu

University of Arizona, Tucson, AZ

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📄 <https://xiaolang.github.io/>

## EDUCATION

- **University of Arizona, Tucson, AZ** 2018 - Present  
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

- **Research Intern, Security Lab, Baidu Inc. USA** Summer 2019  
Mentor: Dr. Yueqiang Cheng Sunnyvale, CA
  - ◇ Job Description:
    - Contribute to breakthrough innovations in technologies of security-oriented big data analysis.
    - For private key-value data collection, we developed a novel framework with an advanced data sampling method and optimized perturbation of key-value pairs, which outperforms state-of-the-art protocol.
    - Publication highlight: top-tier conference (refer to paper [USENIX Security' 20])
- **Graduate Research Assistant, University of Arizona** Fall 2018 - Present  
Advisor: Dr. Ming Li Tucson, AZ
  - ◇ Local Differential Privacy and Its Applications (publication highlight: top-tier conferences)
    - To simultaneously enhance the utility for record-level queries and statistical/aggregated analysis, we proposed a novel and practical protocol for location-based applications (refer to paper [CNS' 19])
    - For private frequency estimation of categorical data, we proposed a relaxed privacy notion that provides differentiated privacy guarantees for data inputs that have distinct privacy requirements, and developed a practical protocol with optimized parameters to get the benefit from the non-uniform indistinguishability (refer to paper [ICDE' 20])
  - ◇ Differentially Private and Robust Machine Learning/Federated Learning
    - Compared with centralized setting, federated learning with differential privacy (DP) suffers from bad privacy-utility tradeoff and is vulnerable to model poisoning attacks. To mitigate these challenges, we leverage secure multiparty computation techniques and develop a hybrid solution (with both DP and crypto), which achieves better privacy-utility tradeoff and provides robustness guarantee against model poisoning attacks.

## SKILLS

- Python (numpy, scipy, pandas, pytorch, sklearn), Matlab, C
- Privacy-preserving techniques: differential privacy (DP) and secure multiparty computation (SMC)
- Robust machine learning/federated learning against adversarial examples and poisoning attacks

## RELEVANT COURSEWORK

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

- Conference Papers

- Journal Papers

- ## PROFESSIONAL SERVICES

- ## AWARDS AND HONORS

- ## TEACHING

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