

# Tao Chen

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## EDUCATION

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### Xi'an University of Posts & Telecommunications

Xian, China

*B.Sc. in Information Security; GPA: 3.58/4.00*

*Sep 2017 – Jul 2021*

### University of Electronic Science and Technology of China

Chengdu, China

*M.Sc. in Cryptography; GPA: 3.45/4*

*Sep 2021 – Present*

## RESEARCH EXPERIENCE

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### Ciphertext Deduplication for Secure Cloud Storage

XUPT, Xian, Shanxi

*Undergraduate Student*

*Feb 2020 – Jul 2021*

- worked in the laboratory of Prof. Meixia Miao with her master students on a project about ciphertext deduplication for cloud storage.
- completed my dissertation "A Lightweight Ciphertext Deduplication Technology with Public Data Auditing" supervised by Prof. Meixia Miao and Dr. Guohua Tian who studied at Xidian University.

### Applied Cryptography & Federated Learning

UESTC, Chengdu, China

*Master Student*

*Sep 2021 – present*

- worked with Prof. Xiaofen Wang on the project "Privacy-preserving and Verifiable Decentralized Federated Learning".
- studied homomorphic encryption consisting of single-key homomorphic encryption and multi-key homomorphic encryption and designed a new notion of homomorphic encryption called multi-key homomorphic encryption with single-key decryption.
- studied identity-based broadcast encryption and designed a new identity-based broadcast encryption algorithm used to provide privacy protection and dropout tolerance for federated learning.
- studied functional encryption consisting of multi-input functional encryption, multi-client functional encryption, and decentralized functional encryption.

## PUBLICATIONS

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### Papers

#### First Author (Accept)

IEEE International Conference on Communications

- Communication-Efficient Privacy-Preserving and Verifiable Federated Learning

#### First Author (Accept)

IEEE Global Communications Conference

- DTPP-DFL: A Dropout-Tolerated Privacy-Preserving Decentralized Federated Learning Framework

#### First Author (Accept)

IEEE Internet of Things Journal

- A Dropout-Tolerated Privacy-Preserving Method for Decentralized Crowdsourced Federated Learning

#### First Author (Manuscript in preparation)

IEEE Transactions on Information Forensics and Security

- An Efficient Privacy-Preserving and Verifiable Federated Learning Framework based on Multi-Key Homomorphic Encryption

#### First Author (Manuscript in preparation)

CT-RSA

- Decentralized Mult-client Functional Encryption without private channels

#### Sixth Author (Under Review)

Peer-to-Peer Networking and Applications (PPNA)

- An Efficient Blockchain-Based Self-Tallying Voting Protocol with Full-Anonymity

#### Third Author (Accept)

The Sixth International Conference on Frontiers in Cyber Security

- ASEV: Anonymous and Scored-Based E-Voting Protocol on Blockchain

### Patents

- A Verifiable Privacy-Preserving Federated Learning Method and System.
- A Certificateless Anonymous Cross-Domain Authentication Method for IoT devices based on Blockchain.
- An Efficient Fully Anonymous Self-Counting Voting Method Based on Blockchain.

## PROJECTS

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### Programming Projects

- Multi-receiver anonymous signcryption for crowdsourced IoMT** | [GitHub](#) Finished
- A python project which implements a multi-receiver anonymous signcryption algorithm used for trusted health data collection in Internet-of-Medical-Things and reproduces an existing multi-receiver signcryption scheme.
  - The multi-receiver anonymous signcryption integrates the attribute-based credential (ABC) and the identity-based multi-receiver encryption technologies, which can achieve fine-grained, privacy-preserving authentication and satisfy the necessary properties of traceability and non-repudiation.
- A privacy-preserving and verifiable federated learning** | [GitHub](#) Finished
- A python project which implements a privacy-preserving and verifiable federated learning framework, this project consists the following aspects: 1) a novel multi-key homomorphic encryption called multi-key homomorphic encryption with single-key decryption used for providing privacy protection for federated learning; 2) a vector homomorphic hash function used for federated learning, which can ensure the integrity of the local model and global model; 3) A reproduction of two existing traditional multi-key homomorphic encryption schemes used for federated learning and a comparison of overhead between these two schemes and the novel multi-key homomorphic encryption scheme.
- A reproduction of poisoning attacks and defence in federated learning** | [GitHub](#) Finished
- A python project which implements some poisoning attacks, such as label-flipping attacks and model replacing attacks and the corresponding defense strategies, such as Krum, Multi-Krum, and Bulyan.
- A novel ID-based homomorphic broadcast encryption & Federated learning** | [GitHub](#) Finished
- This project consists of a novel identity-based homomorphic broadcast encryption algorithm with aggregation decryption used for federated learning, which can provide privacy protection. Besides, the new identity-based broadcast encryption can support dropout tolerance, which means that no matter how many clients drop out from the federated learning system, it can still correctly provide privacy protection.

### Writing Projects

- A fund declaration about a decentralized secure federated learning framework.** Finished
- This project is a declaration of the National Natural Science Foundation project, which has already been granted, and the main content is about a blockchain-based privacy-preserving and verifiable decentralized federated learning framework for medical treatment. The serial number is 62372092.
- A brochure of a Go homomorphic encryption library Lattigo.** | [GitHub](#) Ongoing
- In this project, I'm writing a brochure for a homomorphic encryption library called Lattigo, which includes explanations about how to use these functions and the rationales for some functions in Lattigo. The brochure is in Chinese now but I will translate it to English in the future.

## SKILLS

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**Programming:** C, C++, Python, MATLAB, Go

**Cryptographic Libraries:**(Python) Crypto, Pyseal, Pypbc, Gmpy2; (C& C++): SEAL, Relic, GMP, NTL, CiFEr, Miracle; (Go) Lattigo

**Languages:** Chinese (Native), English (IELTS 6.0)

## RELEVANT COURSEWORK

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**Major coursework:**Modern Cryptography, Machine Learning, Number Theory, Abstract Algebra, Computer Network, Data Structures and Algorithms, Graph Theory

**Minor coursework:**Theory of Combinatorial Design and Optimization, Internet Security, Cloud Computing, The general theory of information security