

Xu, Jianyu

xu_jy15@ucsb.edu

(+1)805-886-7689

xu-jy.github.io/

Department of Computer Science
University of California, Santa Barbara
CA, 93106

EDUCATION

- 2019.9-current **PhD student in Computer Science, University of California at Santa Barbara**
 Advisor: Prof. Yu-Xiang Wang, and Prof. Zheng Zhang
 GPA: **3.94/4.0**
- 2015.8-2019.7 **B.S. in Measurement and Control Technology and Instrument, Tsinghua University, China**
 Advisor: Prof. Guoqi Li
 GPA: **3.74/4.0** Rank: 4/59

RESEARCH INTERESTS

Currently I am working on *Dynamic Pricing* problems. My interest mainly lies broadly on **statistical machine learning** and **online decision-making**.

In the past few years, I have also been working in the following fields:

- Graph Theory, Computational Complexity, and Combinatorics
- Tensor Network and Calculus

AWARDS AND HONORS

- 2022 NeurIPS 2022 Reviewer Award (Top 8%)
- 2019 Excellent Undergraduate Student Award, Tsinghua University (Top 5%)
- 2018 Recommendation, by Department of PI, for **Special Scholarship** of Tsinghua University
- 2014 Silver Medal, 30th Chinese Mathematical Olympiad (CMO)
- 2014 First Prize and **Provincial Champion** (1st /20,000+), Chinese High School Mathematical Contest
- 2013 Silver Medal, 29th Chinese Mathematical Olympiad (CMO)
- 2013 First Prize, National Senior High School Mathematical Contest

PUBLICATIONS [[Google Scholar](#)]

(* for equal contributions.)

Preprint and Working Papers:

- **Jianyu Xu**, Dan Qiao, and Yu-Xiang Wang, “Doubly Fair Dynamic Pricing.” arXiv preprint arXiv: 2209.11837.
- **Xu, Jianyu**, Hanwen Zhang, Lei Deng, and Guoqi Li. “NP-hardness of tensor network contraction ordering.” (*working paper*).

Conference Papers:

- **Xu, Jianyu**, and Yu-Xiang Wang, "Towards Agnostic Feature-based Dynamic Pricing: Linear Policies vs Linear Valuation with Unknown Noise." in *AISTATS 2022*. (**Plenary Oral Presentation, <3%**)
- **Xu, Jianyu**, and Yu-Xiang Wang, "Logarithmic Regret in Feature-based Dynamic Pricing." in *NeurIPS 2021*. (**Spotlight Presentation, <3%**)

Journal Papers:

- Dheeraj Baby*, **Jianyu Xu***, and Yu-Xiang Wang, "Non-stationary Contextual Pricing with Safety Constraints." Accepted by *Transactions on Machine Learning Research*, 2022.
- Liang, Ling, **Jianyu Xu**, Lei Deng, Mingyu Yan, Xing Hu, Zheng Zhang, Guoqi Li, and Yuan Xie. "Fast Search of the Optimal Contraction Sequence in Tensor Networks." *IEEE Journal of Selected Topics in Signal Processing* 15, no. 3 (2021): 574-586. (**Cover Paper**)
- **Xu, Jianyu**, Ling Liang, Lei Deng, Changyun Wen, Yuan Xie, and Guoqi Li. "Towards a polynomial algorithm for optimal contraction sequence of tensor networks from trees." *Physical Review E* 100, no. 4 (2019): 043309.
- **Xu, Jianyu**, Guoqi Li, Changyun Wen, Kun Wu, and Lei Deng. "Towards a unified framework of matrix derivatives." *IEEE Access* 6 (2018): 47922-47934.

INTERNSHIP

2022.06 – 2022.09 Applied Scientist Intern at Amazon, Seattle

In Retail Pricing Science & Research Team,

Supervised by Dr. Pau Pereira, hosted by Dr. Tara Mardan

- Develop multi-armed bandit algorithms for Amazon retail pricing systems to escalate long-term revenue.
- Apply Fourier Transformation to simulate real-world demand-to-price data for algorithm testings.

2021.07 – 2021.10 Research Intern at AntGroup, Beijing & Hangzhou

Supervised by Dr. Wenpeng Zhang

- Develop algorithms on attracting new/sleeping/lost customers with personalized-value coupons.
- Study “contextual bandits with knapsacks” for budget-constraint coupon pricing.

RESEARCH EXPERIENCE

2019.11 – current Decision Making and Dynamic Pricing

Advised by Prof. Yu-Xiang Wang, Dept. Computer Science, UCSB

- Develop algorithms for online dynamic pricing under different assumptions.
- Prove regret upper & lower bounds for these algorithms.

2018.1 – 2019.8 NP-Hardness of Tensor Network Contraction Ordering

Advised by Prof. Guoqi Li, Department of Precision Instrument, Tsinghua University

and Prof. Yuan Xie, Scalable Energy-Efficient Architecture Lab, UCSB (2018.7-2018.9)

- Given the existing problem setting to be NP-hard, propose an easier version of the problem setting.
- Prove the easiness: by pointing out a case which is polynomial in the new version, but NP-hard in the old.
- Prove the hardness: even the easier version is also NP-hard.

2017.2– 2018.2 Computation on Matrix Function Derivatives

Advised by Prof. Guoqi Li, Department of Precision Instrument, Tsinghua University

- Conclude 2 main approaches of calculating matrix-to-scalar function derivatives in chain rule.
- Proved their equivalence under certain conditions.

TEACHING ASSISTANTSHIP

2020 Spring	CS 165A, <i>Artificial Intelligence</i> , Dept. CS, UCSB
2020 Winter	CS 165A, <i>Artificial Intelligence</i> , Dept. CS, UCSB
2019 Fall	CS 8, <i>Introduction to Computer Science</i> , Dept. CS, UCSB