

Tengyuan Liang

curriculum vitae

5807 South Woodlawn Avenue
Chicago, IL 60637
☎ +1 773.702.7935

✉ Tengyuan.Liang@ChicagoBooth.edu
📄 tyliang.github.io/Tengyuan.Liang/

Academic Appointment

2017– **University of Chicago, Booth School of Business**, United States.
Associate Professor of Econometrics and Statistics (without Tenure), 2021.07 –
Assistant Professor of Econometrics and Statistics, 2017.07 – 2021.06

Becker Friedman Institute, *Big Data Initiative*.
Affiliated Scholar, 2018 –

Education

2012–2017 **University of Pennsylvania, The Wharton School**, United States.
Ph.D. in Statistics

2008–2012 **Peking University**, China.
B.S. in Mathematics and Applied Mathematics

Visiting Positions

2019 Yale University, *Cowles Foundation for Research in Economics*.
Visiting Assistant Professor in Econometrics

2016 Yahoo Research New York, *Online Learning and Optimization Group*.
Summer Research Scientist

Honors & Awards

2021–2026 NSF Career Award
[DMS - 2042473 "New Statistical Paradigms Reconciling Empirical Surprises in Modern Machine Learning"](#), by
National Science Foundation, Division of Mathematical Sciences

2021–2022 William S. Fishman Faculty Scholar

2017–2021 George C. Tiao Faculty Fellow
[research fellowship for computational and data science](#) awarded by the Booth School

2014–2017 Winkelman Fellowship
[highest honorific fellowship](#) awarded by the Wharton School

2016 J. Parker Memorial Bursk Award
[awarded by the Statistics Department at the Wharton School](#) for excellence in research

2014 US Junior Oberwolfach Fellow

Research

Research Interests

Fields: Learning Theory, Mathematical Statistics, Econometrics.

Working Papers

- 24. W. Guo, Y. Hur and T. Liang (2021). wp
"Reversible Gromov-Monge Sampler."
working paper, available upon request

- 23. T. Liang (2021). arXiv
"Universal Prediction Band via Semi-Definite Programming."
arXiv:2103.17203 [v1]

- 22. T. Liang and B. Recht (2021). arXiv
"Interpolating Classifiers Make Few Mistakes."
arXiv:2101.11815 [v2]

- 21. M. H. Farrell, T. Liang and S. Misra (2020). arXiv
"Deep Learning for Individual Heterogeneity: An Automatic Inference Framework."
arXiv:2010.14694 [v2]

- 20. T. Liang and P. Sur (2020). arXiv
"A Precise High-Dimensional Asymptotic Theory for Boosting and Minimum- ℓ_1 -Norm
Interpolated Classifiers."
arXiv:2002.01586 [v3] , *Ann. Stat.*, *R&R*

- 19. T. Liang (2019). arXiv
"Estimating Certain Integral Probability Metrics (IPMs) Is as Hard as Estimating under
the IPMs."
arXiv:1911.00730 [v1]

- 18. T. Liang (2018). arXiv
"How Well Generative Adversarial Networks Learn Distributions."
arXiv:1811.03179 [v4] , *J. Mach. Learn. Res.*, *R&R*

Publications

- 17. T. Liang and H. Tran-Bach (2020). JASA
"Mehler's Formula, Branching Process, and Compositional Kernels of Deep Neural
Networks."
Journal of the American Statistical Association (Theory and Methods), 1-14, 2021.

- 16. M. H. Farrell, T. Liang and S. Misra (2018). ECMA
"Deep Neural Networks for Estimation and Inference."
Econometrica, 89(1):181-213, 2021.

- 15. T. Liang, A. Rakhlin and X. Zhai (2019). COLT
"On the Multiple Descent of Minimum-Norm Interpolants and Restricted Lower Isome-
try of Kernels."
Conference on Learning Theory, pmlr 125:2683-2711, 2020.

- 14. X. Dou and T. Liang (2019). JASA
"Training Neural Networks as Learning Data-adaptive Kernels: Provable Representation
and Approximation Benefits."
Journal of the American Statistical Association (Theory and Methods), 1-14, 2020.

- 13. T. Liang and A. Rakhlin (2018). AOS
"Just Interpolate: Kernel "Ridgeless" Regression Can Generalize."
The Annals of Statistics, 48(3):1329-1347, 2020.

12. T. Liang and W. J. Su (2017). JRSS-B
 "Statistical Inference for the Population Landscape via Moment Adjusted Stochastic Gradients."
Journal of the Royal Statistical Society: Series B (Statistical Methodology), 81(2):431-456, 2019.
11. T. T. Cai, T. Liang and A. Rakhlin (2017). JMLR
 "Weighted Message Passing and Minimum Energy Flow for Heterogeneous Stochastic Block Models with Side Information."
Journal of Machine Learning Research, 21(11):1-34, 2020.
10. T. Liang and J. Stokes (2018). AISTATS
 "Interaction Matters: A Note on Non-asymptotic Local Convergence of Generative Adversarial Networks."
International Conference on Artificial Intelligence and Statistics, pmlr 89:907-915, 2019.
9. T. Liang, T. Poggio, A. Rakhlin and J. Stokes (2017). AISTATS
 "Fisher-Rao Metric, Geometry, and Complexity of Neural Networks."
International Conference on Artificial Intelligence and Statistics, pmlr 89:888-896, 2019.
8. B. Tzen, T. Liang and M. Raginsky (2018). COLT
 "Local Optimality and Generalization Guarantees for the Langevin Algorithm via Empirical Metastability."
Conference on Learning Theory, pmlr 75:857-875, 2018.
7. S. Kale, Z. Karnin, T. Liang and D. Pál (2017). ICML
 "Adaptive Feature Selection: Computationally Efficient Online Sparse Linear Regression under RIP."
International Conference on Machine Learning, pmlr 70:1780-1788, 2017.
6. T. T. Cai, T. Liang and A. Rakhlin (2017). AOS
 "Computational and Statistical Boundaries for Submatrix Localization in a Large Noisy Matrix."
The Annals of Statistics, 45(4):1403-1430, 2017.
5. T. T. Cai, T. Liang and A. Rakhlin (2017). IEEE-TNSE
 "On Detection and Structural Reconstruction of Small-World Random Networks."
IEEE Transactions on Network Science and Engineering, 4(3):165-176, 2017.
4. T. T. Cai, T. Liang and A. Rakhlin (2016). AOS
 "Geometric Inference for General High-Dimensional Linear Inverse Problems."
The Annals of Statistics, 44(4):1536-1563, 2016.
3. T. Liang, A. Rakhlin and K. Sridharan (2015). COLT
 "Learning with Square Loss: Localization through Offset Rademacher Complexity."
Conference on Learning Theory, pmlr 40:1260-1285, 2015.
[nominated for the best paper award](#)
2. A. Belloni, T. Liang, H. Narayanan and A. Rakhlin (2015). COLT
 "Escaping the Local Minima via Simulated Annealing: Optimization of Approximately Convex Functions."
Conference on Learning Theory, pmlr 40:240-265, 2015.
1. T. T. Cai, T. Liang and H. H. Zhou (2015). JMVA
 "Law of Log Determinant of Sample Covariance Matrix and Optimal Estimation of Differential Entropy for High-Dimensional Gaussian Distributions."
Journal of Multivariate Analysis, 137:161-172, 2015.

Professional Activities

2020– **Editorial Board**, *Journal of Machine Learning Research*.

2020– **Senior Program Committee**, *Conference on Learning Theory (COLT)*.

2014– **Journal and Conference Referee**.

- **Statistics and Probability**: *Annals of Statistics*, *Journal of the Royal Statistical Society Series B (Statistical Methodology)*, *Journal of the American Statistical Association (Theory and Methods)*, *Biometrika*, *Bernoulli Journal*, *Statistica Sinica*, *Latin American Journal of Probability and Mathematical Statistics*, *Statistical Science*, *Probability Theory and Related Fields*.
- **Learning Theory**: *Journal of Machine Learning Research*, *Conference on Learning Theory (COLT)*, *Symposium on the Theory of Computing (STOC)*, *International Conference for Learning Representations (ICLR)*.
- **Econometrics**: *Econometrica*, *Journal of Econometrics*, *Review of Economics and Statistics*.
- **Information Theory**: *IEEE Transactions on Information Theory*, *IEEE International Symposium on Information Theory (ISIT)*.
- **Operations Research**: *Mathematics of Operations Research*.
- **Applied Mathematics**: *SIAM Journal on Mathematics of Data Science*.

Invited Presentations

- 2021–2022 [scheduled] ○ UBC [Operations Research Seminar, Sauder School of Business]
○ UCL [Econometrics Seminar, Dept. of Economics]
○ UIUC [Statistics Seminar, Dept. of Statistics]
- 2020–2021 ○ NSF-Simons Research Collaborations [Mathematics of Deep Learning Workshop, 60 mins talk]
○ UMass Amherst [Statistics and Probability Seminar, Dept. of Mathematics and Statistics]
○ Rutgers [Statistics Seminar, Dept. of Statistics]
○ Durham [Seminar at Business School]
○ LSE [Econometrics Seminar, Dept. of Economics]
- 2019–2020 ○ MIT [Statistics and Stochastics Seminar Series, IDSS]
○ Yale [Econometrics Seminar, Dept. of Economics]
○ Harvard [Statistics Colloquium, Dept. of Statistics]
○ MIT [MIFODS Workshop “Learning with a complex structure,” 45 mins talk]
○ Duke [TRIPODS Workshop “Theory and modeling of deep learning,” 50 mins talk]
○ Google Research NYC [Learning Theory Seminar]
- 2018–2019 ○ Duke [Decision Sciences Seminar, Fuqua School of Business]
○ ENSAE-CREST [Center for Research in Economics and Statistics Seminar]
○ UChicago [Joint U Chicago and TTIC Machine Learning Seminar]
- 2017–2018 ○ UIUC [Machine Learning Seminar, ECE Dept.]
○ UChicago [Statistics Colloquium, Dept. of Statistics]
○ HKUST [Joint Statistics Seminar, Business School and Dept. of Mathematics]

- 2016–2017
- Stanford [Statistics Seminar, Dept. of Statistics]
 - Princeton [Colloquia, Operation Research and Financial Engineering]
 - MIT [Operations Research and Statistics Seminar, Sloan School of Management]
 - UChicago [Econometrics and Statistics Seminar, Booth School of Business]
 - Cambridge [Statistical Laboratory Seminar, Dept. of Mathematics]
 - Georgia Tech [Statistics Seminar, Dept. of Mathematics]
 - UCSD [Statistics Seminar, Dept. of Mathematics]
 - UVA [Statistics Seminar, Dept. of Statistics]
 - UIUC [Statistics Seminar, Dept. of Statistics]
 - Rutgers [Statistics Seminar, Dept. of Statistics, cancelled]
 - Imperial College London [Operations Management Seminar, Business School]
 - Yahoo Labs [Machine Learning Seminar]

Conferences ICML 2021 [Invited Speaker, Workshop “Over-parameterization: Pitfalls and Opportunities”], JSM 2020 [IMS Invited Session, “Theory of Deep Learning”], ICCOPT 2019 [Generalization and Optimization Invited Session], JSM 2019 [Invited Session on “Modern Nonparametrics”], AISTATS 2019 [Present Two Papers], DALI 2019 [Machine Learning Theory Invited Session, George, South Africa], Econometric Conference on Big Data [Invited Talk at “Factor Models” Session, Tsinghua Univ.], COLT 2018 [Stochastic Optimization Session, KTH], Issac Newton Institute [Workshop on Future Challenges in Statistical Scalability, Cambridge], EcoSta 2018 [Frontiers in Financial Statistics Invited Session, CityU Hong Kong], CISS 2018 [Statistical Learning Invited Session, Princeton], ICML 2017 [Online Learning Session, Sydney], COLT 2015 [University Pierre and Marie Curie, Two Long Talks], Yale [NSF Workshop for Empirical Process and Modern Statistical Decision Theory], CIRM [Meeting in Mathematical Statistics: New Procedures for New Data, Luminy, France], CRM [Workshop on the Mathematical Foundations of Learning Theory, Barcelona, Spain], MFO [Workshop on Adaptive Statistical Inference, Oberwolfach, Germany].

Teaching Experience

- 2017– **University of Chicago Booth School of Business, Instructor.**
- Spring 18, Fall 18, Fall 19, Fall 20, Fall 21: Business Statistics [BUSN41000, MBA]
- 2012–2017 **Wharton School at University of Pennsylvania, Teaching Assistant.**
- Spring 17: Stochastic Processes [STAT931, PhD]
 - Spring 15: Concentration Inequalities [STAT991, PhD]
 - Spring 14, Spring 15: Advanced Quantitative Modeling [STAT622, MBA]
 - Fall 13: Introductory Statistics [STAT111]
 - Spring 13: Statistical Inference [STAT431]
 - Fall 12, Fall 15, Fall 16: Probability [STAT430]

University Service

- 2020– **Organizer, Econometrics and Statistics Colloquium, Chicago Booth.**

Mentoring & Advising

- 2018– **Doctoral Students.**
- Wenxuan Guo [2026, PhD, Chicago Booth], YoonHaeng Hur [2024, PhD, UChicago Stat], Hai Tran-Bach [2023, PhD, UChicago Stat], Xialiang Dou [2021, PhD, UChicago Stat]
- 2018– **Dissertation Committee.**
- Sen Na [2021, PhD, UChicago Stat → Postdoc, Berkeley], Shihao Gu [2021, PhD, Chicago Booth], Jingyu He [2020, PhD, Chicago Booth → Assistant Professor, CityU Hong Kong], Ming Yu [2020, PhD, Chicago Booth → Citadel], Qi An [2019, PhD, Chicago Booth]
- 2018– **Organizer, Data Science Reading Group, UChicago.**
- mentoring PhD students from Statistics and Computational and Applied Mathematics [list of papers]