









# RAAZ DWIVEDI

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## ACADEMIC APPOINTMENTS

**Postdoctoral Fellow**, Computer Science & Statistics, Harvard University and Electrical Engineering (EE) & Computer Sciences (CS), Massachusetts Institute of Technology (MIT) 2021–  
Advisors: *Prof. Susan Murphy & Prof. Devavrat Shah*

## EDUCATION

**Ph.D., EECS**, University of California (UC), Berkeley 2015–2021  
Advisors: *Prof. Martin Wainwright & Prof. Bin Yu*  
Thesis title: *Principled statistical approaches for sampling and inference in high dimensions*  
**B. Tech., EE**, Indian Institute of Technology (IIT), Bombay, India 2010–2014  
Minors in mathematics, Institute Rank 1

## RESEARCH INTERESTS

My research involves a multi-disciplinary approach to data science and brings together ideas from computer science, electrical engineering, and statistics in collaboration with domain experts. I develop statistical machine learning approaches for data-driven personalized decision-making with research across *causal inference, reinforcement learning, Bayesian inference, random sampling, and high-dimensional statistics*.

## SELECTED ACHIEVEMENTS & AWARDS

Best Student Paper Award, Statistical Computing & Graphics, American Statistical Association 2022  
Best Presentation Award, Laboratory of Information and Decision Systems (LIDS) Conference, MIT 2022  
Certificate of Distinction and Excellence in Teaching (Q Award), Harvard University 2022  
Foundations of Data Science (FODSI) Postdoctoral Fellowship 2021  
Outstanding Graduate Student Instructor Award, UC Berkeley 2020  
Berkeley Fellowship, the most prestigious fellowship for incoming Ph. D. students 2015  
President of India Gold Medal, IIT Bombay, for the highest GPA in the graduating class 2014  
All India Rank 10 amongst half a million, IIT Joint Entrance Exam 2010

## CONFERENCE PUBLICATIONS

- ( $\star$  denotes equal contribution and  $\dagger$  denotes alphabetical ordering; title is hyperlinked to the online pdf of the paper)
- C1. **Raaz Dwivedi**, Lester Mackey. “Generalized kernel thinning”, *International Conference on Learning Representations (ICLR)*. 2022
  - C2. Abhishek Shetty, **Raaz Dwivedi**, Lester Mackey. “Distribution compression in near-linear time”, *International Conference on Learning Representations (ICLR)*, **Best Student Paper Award, JSM**. 2022
  - C3. **Raaz Dwivedi**, Lester Mackey, “Kernel thinning”, Extended abstract in *Conference on Learning Theory (COLT)*. Full version under review in *JMLR*. 2021
  - C4. **Raaz Dwivedi** $\star$ , Nhat Ho $\star$ , Koulik Khamaru $\star$ , Martin J. Wainwright, Michael I. Jordan, Bin Yu, “Sharp analysis of Expectation-Maximization for weakly identifiable models”, *The 23rd International Conference on Artificial Intelligence and Statistics (AISTATS)*. 2020
  - C5. **Raaz Dwivedi** $\star$ , Nhat Ho $\star$ , Koulik Khamaru $\star$ , Martin J. Wainwright, Michael I. Jordan, “Theoretical guarantees for EM under misspecified Gaussian mixture models”, *Advances in Neural Information Processing Systems (NeurIPS)*. 2018
  - C6. **Raaz Dwivedi** $\star$ , Yuansi Chen $\star$ , Martin J. Wainwright, Bin Yu, “Log-concave sampling: Metropolis-Hastings algorithms are fast”, Extended abstract in *Conference on Learning Theory (COLT)*. 2018
  - C7. Yuansi Chen $\star$ , **Raaz Dwivedi** $\star$ , Martin J. Wainwright, Bin Yu, “Vaidya walk: A sampling algorithm based on the volumetric barrier”, *Allerton Conference*. 2017
  - C8. **Raaz Dwivedi**, Vivek Borkar, “Removing sampling bias in networked stochastic approximation”, *International Conference on Signal Processing and Communications (SPCOM)*. 2014

JOURNAL  
PUBLICATIONS

- J1. Nhat Ho<sup>\*</sup>, Koulik Khamaru<sup>\*</sup>, **Raaz Dwivedi**<sup>\*</sup>, Martin J. Wainwright, Michael I. Jordan, Bin Yu, “Instability, computational efficiency, and statistical accuracy”, *accepted with minor revision at JMLR*. 2022+
- J2. **Raaz Dwivedi**<sup>\*</sup>, Chandan Singh<sup>\*</sup>, Bin Yu, Martin J. Wainwright, “Revisiting minimum description length complexity in overparameterized models”, *accepted with minor revision at JMLR*. 2022+
- J3. Nick Altieri<sup>†</sup>, Rebecca L. Barter, James Duncan, **Raaz Dwivedi**, Karl Kumbier, Xiao Li, Robert Netzorg, Briton Park, Chandan Singh, Yan Shuo Tan, Tiffany Tang, Yu Wang, Chao Zhang, Bin Yu, “Curating a COVID-19 data repository and forecasting county-level death counts in the United States”, *Harvard Data Science Review (HDSR)*. 2021
- J4. **Raaz Dwivedi**<sup>\*</sup>, Yan Shuo Tan<sup>\*</sup>, Briton Park, Mian Wei, Kevin Horgan, David Madigan, Bin Yu, “Stable discovery of interpretable subgroups via calibration in causal studies”, *Int. Statistical Review*. 2020
- J5. **Raaz Dwivedi**<sup>\*</sup>, Nhat Ho<sup>\*</sup>, Koulik Khamaru<sup>\*</sup>, Martin J. Wainwright, Michael I. Jordan, Bin Yu, “Singularity, misspecification, and the convergence rate of EM”, *Annals of Statistics (AoS)*. 2020
- J6. Yuansi Chen, **Raaz Dwivedi**, Martin J. Wainwright, Bin Yu, “Fast mixing of Metropolized Hamiltonian Monte Carlo: Benefits of multi-step gradients”, *Journal of Machine Learning Research (JMLR)*. 2020
- J7. **Raaz Dwivedi**<sup>\*</sup>, Yuansi Chen<sup>\*</sup>, Martin J. Wainwright, Bin Yu, “Log-concave sampling: Metropolis-Hastings algorithms are fast”, *Journal of Machine Learning Research (JMLR)*. 2019
- J8. **Raaz Dwivedi**<sup>†</sup>, Ohad N. Feldheim, Ori Gurel-Gurevich, Aaditya Ramdas. “The power of online thinning in reducing discrepancy”, *Probability Theory and Related Fields (PTRF)*. 2019
- J9. Yuansi Chen<sup>\*</sup>, **Raaz Dwivedi**<sup>\*</sup>, Martin J. Wainwright, Bin Yu. “Fast MCMC sampling algorithms on polytopes”, *Journal of Machine Learning Research (JMLR)*. 2018
- J10. Vivek Borkar<sup>†</sup>, **Raaz Dwivedi**, Neeraja Sahasrabudhe. “Gaussian approximations in high dimensional estimation”, *Systems & Control Letters*. 2016

PRE-PRINTS &  
WORKING PAPERS

- P1. **Raaz Dwivedi**, Katherine Tian, Sabina Tomkins, Predrag Klasnja, Susan Murphy, Devavrat Shah, “Counterfactual inference in sequential experimental design”, *arxiv, to be submitted to Annals of Statistics (AoS)*. 2022
- P2. **Raaz Dwivedi**, Katherine Tian, Sabina Tomkins, Predrag Klasnja, Susan Murphy, Devavrat Shah, “Doubly robust nearest neighbors in factor models”, *preliminary version on arxiv*. 2022
- P3. Abhin Shah, **Raaz Dwivedi**, Devavrat Shah, Greg Wornell, “On counterfactual inference with unobserved confounding”, *NeurIPS workshop, full version to be submitted to AoS*. 2022
- P4. Carles Domingo-Enrich, **Raaz Dwivedi**, Lester Mackey, “Compress then test: Powerful kernel testing in near-linear time”, *in conference submission*. 2022
- P5. **Raaz Dwivedi**<sup>\*</sup>, Kelly Zhang<sup>\*</sup>, Prasidh Chhabria, Predrag Klasnja, Susan Murphy, “Assessing personalization by a reinforcement learning algorithm”, *Working paper*. 2022+

SOFTWARES &  
METHODOLOGIES

- S1. Carles Domingo-Enrich, **Raaz Dwivedi**, Lester Mackey. Python package “Compress then test” ([🔗 link](#)).
- S2. Abhishek Shetty<sup>\*</sup>, **Raaz Dwivedi**<sup>\*</sup>, Lester Mackey. Python package “Compress++” ([🔗 link](#)).
- S3. **Raaz Dwivedi**, Lester Mackey. Python package “Kernel Thinning” ([🔗 link](#)).
- S4. **Raaz Dwivedi**<sup>\*</sup>, Yan Shuo Tan<sup>\*</sup>, Briton Park, Mian Wei, Kevin Horgan, David Madigan, Bin Yu. Python repository “StaDISC” ([🔗 link](#)).
- S5. Yuansi Chen<sup>\*</sup>, **Raaz Dwivedi**<sup>\*</sup>, Martin Wainwright, Bin Yu. Python package (with C++ implementation) “Vaidya and John walks” ([🔗 link](#)).

SELECTED INVITED  
TALKS

**From HeartSteps to HeartBeats: Personalized Decision-making**

- Statistics Seminar, Rutgers University *Nov 2022*
- ISL Colloquium, EE, Stanford University *Nov 2022*
- BLISS Seminar, EECS, UC Berkeley *Nov 2022*

**Counterfactual inference in sequential experiments**

- Informs Annual Meeting, Indianapolis *Oct 2022*
- Institute of Mathematical Statistics (IMS) Annual Meeting, London *Jun 2022*
- Learning from Interventions Workshop, Simons Institute, Berkeley *Feb 2022*

**Near-optimal compression in near-linear time**

- SIAM Conference on Uncertainty Quantification, Atlanta *Apr 2022*
- Statistical learning Workshop, Mathematical Sciences Research Institute, Berkeley *Mar 2022*

**Kernel thinning**

- Data-Centric Engineering Group, Alan Turing Institute, Virtual *Sep 2021*

**Revisiting minimum description length complexity in overparameterized models**

- Alg. Info Theory & Machine Learning Symp., Alan Turing Institute, London *Jul 2022*
- Collaborations on the Theoretical Foundations of Deep Learning, Virtual *Nov 2021*

**StaDISC: Stable discovery of interpretable subgroups via calibration**

- Young Data Scientist Research Seminar, ETH Zurich, Virtual *Sep 2020*
- ASA Annual Symposium on Data Science & Statistics, Virtual *Jun 2020*

**Singularity, misspecification, & the convergence rate of EM**

- Math & Statistics Seminar, IIT Kanpur *Jan 2020*
- AMS Special Sections Meeting, UC Riverside *Nov 2019*

**Theoretical guarantees for MCMC algorithms**

- BIDS Seminar, UC Berkeley *Mar 2019*
- EE Seminar, IIT Bombay *Jan 2018*
- STCS Seminar, TIFR Bombay *Jan 2018*

CONTRIBUTED  
& OTHER RESEARCH  
TALKS

**Counterfactual inference in sequential experiments**

- Statistics and data science conference (SDSCON), MIT *Apr 2022*
- Econometrics Lunch, MIT *Mar 2022*

**Near-optimal compression in near-linear time**

- LIDS Student Conference, MIT, **Best presentation award** *Jan 2022*

**Generalized kernel thinning**

- Joint Statistical Meeting (JSM), Washington DC *Aug 2022*

**Kernel thinning**

- Monte Carlo Methods & Applications (MCM), Virtual *Sep 2021*
- International Society for Bayesian Analysis (ISBA) World Meeting, Virtual *Aug 2021*
- The Bayesian Young Statisticians Meeting (BAYSM), Virtual *Aug 2021*
- Joint Statistical Meeting (JSM), Virtual *Aug 2021*
- Conference on Learning Theory (COLT), Virtual *Aug 2021*
- Subset Selection, International Conference on Machine Learning (ICML), Virtual *Jul 2021*

**Revisiting complexity and the bias-variance tradeoff: Using minimum description length**

- Theory of Overparameterized Machine Learning (TOPML) Workshop, Virtual *Apr 2021*

**Converging fast and slow: Statistics vs optimization**

- BAIR and BDD Retreat, Berkeley, Virtual *Aug 2020*

	<b>Log-concave sampling: Metropolis Hastings algorithms are fast</b> <ul style="list-style-type: none"> <li>Joint Statistical Meeting (JSM), Washington DC</li> </ul>	Dec 2018
	<b>Vaidya walk: A sampling algorithm based on the volumetric barrier</b> <ul style="list-style-type: none"> <li>Allerton Conference</li> </ul>	Oct 2017
CONTRIBUTED POSTER PRESENTATIONS	<b>On counterfactual inference with unobserved confounding</b> <ul style="list-style-type: none"> <li>NeurIPS. Causality for Real world impact workshop, New Orleans</li> </ul>	Nov 2022
	<b>Counterfactual inference in sequential experiments</b> <ul style="list-style-type: none"> <li>Cornell ORIE Young Researchers Workshop, Ithaca</li> <li>Royal Statistical Society (RSS) Conference, Aberdeen, Scotland</li> <li>Synthetic Control Methods Workshop, Data X, Princeton University</li> <li>American Causal Inference Conference (ACIC), UC Berkeley</li> <li>Symposium for Mathematical Sciences (SMaSH), Harvard University</li> <li>Statistics and data science conference (SDSCON), MIT</li> </ul>	Oct 2022 Sep 2022 Jun 2022 May 2022 May 2022 Apr 2022
	<b>Near-optimal compression in near-linear time</b> <ul style="list-style-type: none"> <li>Royal Statistical Society (RSS) Conference, Aberdeen, Scotland</li> </ul>	Sep 2022
	<b>Generalized kernel thinning</b> <ul style="list-style-type: none"> <li>Advances in Approximate Bayesian Inference (AABI), Virtual</li> </ul>	Feb 2022
	<b>Revisiting minimum description length complexity in overparameterized models</b> <ul style="list-style-type: none"> <li>North American School of Information Theory (NASIT), Virtual</li> </ul>	Jun 2021
	<b>Theoretical guarantees for EM under misspecified Gaussian mixture models</b> <ul style="list-style-type: none"> <li>Neural Information Processing Systems (NeurIPS), Montréal, Canada</li> </ul>	Dec 2018
	<b>Log-concave sampling: Metropolis Hastings algorithms are fast</b> <ul style="list-style-type: none"> <li>Conference on Learning Theory (COLT), Stockholm, Sweden</li> </ul>	Jul 2018
	<b>On power of two choices in reducing discrepancy</b> <ul style="list-style-type: none"> <li>SAMSI Workshop, Duke University, Raleigh</li> </ul>	Aug 2017
TEACHING EXPERIENCE (TASHIP)	T1. Sequential Decision Making (STAT 234), <i>Harvard University</i> . Gave four guest lectures and supervised several half-semester long research projects. 2022 T2. Modern Statistical Prediction and Machine Learning (STAT 154), <i>UC Berkeley</i> . Gave one guest lecture, helped redesign the class, and developed new discussion sections, homeworks, and exams. 2019 T3. Introduction to Machine Learning (EECS 189), <i>UC Berkeley</i> . Co-head for the content developments in team of 10+ TAs, helped design discussion sections, homeworks, and exams. 2018 T4. Linear Algebra, Calculus, Differential equations (MA 105, 106, 108, 207), <i>IIT Bombay</i> . Taught teaching sections and several voluntary help sessions that were often attended by 200+ students. 2011–2014	
GUEST LECTURES	L1. Regret analysis of posterior sampling (3 lectures, STAT 234), <i>Harvard University</i> Apr 2022 L2. Offline off-policy reinforcement learning (STAT 234) <i>Harvard University</i> . Feb 2022 L3. Revisiting complexity and the bias-variance tradeoff (STAT 212) <i>UC Berkeley</i> . Apr 2021 L4. Introduction to ensemble methods in machine learning (EECS 189), <i>UC Berkeley</i> . Oct 2019 L5. Introduction to boosting methods (STAT 154), <i>UC Berkeley</i> . Apr 2019	

## ACADEMIC SERVICES

### Undergraduate Research Mentoring

- UC Berkeley, One student that led to a co-authored journal publication 2020–2021
- Harvard, Two students with three co-authored submissions in preparation 2022–

### Institutional Mentoring

- MIT Institute for Data, Systems, & Society (IDSS) Postdoc Mentors for two *PhD* students 2022–
- UC Berkeley Artificial Intelligence Research (BAIR) Buddies for two *incoming PhD* students 2020–2021
- UC Berkeley BAIR Mentoring Program for five *undergraduates* 2017–2021
- IIT Bombay Student Mentoring Program (ISMP) for twelve *incoming undergraduates* 2013–2014
- IIT Bombay Academic Mentoring Program (DAMP) for four *sophomores & juniors* 2012–2014
- IIT Bombay Intensive Mentoring Program for thirty *undergraduates* 2012–2013

### Committees

- Member, Committee on Equality and Diversity, IMS 2022–

### Scientific Meetings

- Chair, New Researchers Group Session, IMS Annual Meeting 2022
- Chair, Statistical Machine Learning Session, IMS Annual Meeting 2022
- Mentor, Summer Institute on Just-in-Time Adaptive Interventions via MRTs 2021

### Graduate Admissions

- EECS Graduate Admissions Committee, MIT 2021
- EECS Graduate Admissions Committee, UC Berkeley 2018–2020

### Reviewing Activities

- *Journals*: JMLR, IEEE-IT, JRSSB, Bernoulli, HDSR, Stats & Comp., SIAM, MOR, Jour. of Causal Inference
- *Conferences*: COLT, ICML, AISTATS, NeurIPS, FOCS, STOC, SODA, AAAI

## WORK EXPERIENCE

- Microsoft Research**, Research Intern with Lester Mackey, New England, USA 2019
- Mist Systems, Juniper Networks, Data Science Intern, Cupertino, USA 2017
- WorldQuant Research, Senior Quantitative Researcher, Mumbai, India 2014–2015
- Stanford University**, Research Intern with Prof. Balaji Prabhakar, USA 2013
- Ivy Mobility, Data Science Intern, Chennai, India 2012

## REFERENCES

SUSAN MURPHY  
Professor, CS & Statistics  
Harvard University  
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LESTER MACKEY  
Principal Researcher  
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MARTIN WAINWRIGHT  
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