






# RAAZ DWIVEDI

 Harvard/MIT

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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 & 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

- (★ denotes equal contribution and † 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**★, Nhat Ho★, Koulik Khamaru★, 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**★, Nhat Ho★, Koulik Khamaru★, 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**★, Yuansi Chen★, 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★, **Raaz Dwivedi**★, Martin J. Wainwright, Bin Yu, “Vaidya walk: A sampling algorithm based on the volumetric barrier”, *Allerton Conference*. 2017

JOURNAL  
PUBLICATIONS

- C8. **Raaz Dwivedi**, Vivek Borkar, “Removing sampling bias in networked stochastic approximation”, *International Conference on Signal Processing and Communications (SPCOM)*. 2014
- 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*

	<b>Revisiting complexity and the bias-variance tradeoff: Using minimum description length</b>	
	• Theory of Overparameterized Machine Learning (TOPML) Workshop, Virtual	<i>Apr 2021</i>
	<b>Converging fast and slow: Statistics vs optimization</b>	
	• BAIR and BDD Retreat, Berkeley, Virtual	<i>Aug 2020</i>
	<b>Log-concave sampling: Metropolis Hastings algorithms are fast</b>	
	• Joint Statistical Meeting (JSM), Washington DC	<i>Dec 2018</i>
	<b>Vaidya walk: A sampling algorithm based on the volumetric barrier</b>	
	• Allerton Conference	<i>Oct 2017</i>
CONTRIBUTED POSTER PRESENTATIONS	<b>On counterfactual inference with unobserved confounding</b>	
	• NeurIPS Causality for Real world impact workshop, New Orleans	<i>Nov 2022</i>
	<b>Counterfactual inference in sequential experiments</b>	
	• Cornell ORIE Young Researchers Workshop, Ithaca	<i>Oct 2022</i>
	• Royal Statistical Society (RSS) Conference, Aberdeen, Scotland	<i>Sep 2022</i>
	• Synthetic Control Methods Workshop, Data X, Princeton University	<i>Jun 2022</i>
	• American Causal Inference Conference (ACIC), UC Berkeley	<i>May 2022</i>
	• Symposium for Mathematical Sciences (SMaSH), Harvard University	<i>May 2022</i>
	• Statistics and data science conference (SDSCON), MIT	<i>Apr 2022</i>
	<b>Near-optimal compression in near-linear time</b>	
	• Royal Statistical Society (RSS) Conference, Aberdeen, Scotland	<i>Sep 2022</i>
	<b>Generalized kernel thinning</b>	
	• Advances in Approximate Bayesian Inference (AABI), Virtual	<i>Feb 2022</i>
	<b>Revisiting minimum description length complexity in overparameterized models</b>	
	• North American School of Information Theory (NASIT), Virtual	<i>Jun 2021</i>
	<b>Theoretical guarantees for EM under misspecified Gaussian mixture models</b>	
	• Neural Information Processing Systems (NeurIPS), Montréal, Canada	<i>Dec 2018</i>
	<b>Log-concave sampling: Metropolis Hastings algorithms are fast</b>	
	• Conference on Learning Theory (COLT), Stockholm, Sweden	<i>Jul 2018</i>
	<b>On power of two choices in reducing discrepancy</b>	
	• SAMSI Workshop, Duke University, Raleigh	<i>Aug 2017</i>
TEACHING (TASHIP) EXPERIENCE	T1. Sequential Decision Making (STAT 234), <i>Harvard University</i> . Gave four guest lectures and supervised several half-semester long research projects.	<i>2022</i>
	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.	<i>2019</i>
	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.	<i>2018</i>
	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.	<i>2011–2014</i>

GUEST LECTURES	L1. Regret analysis of posterior sampling (3 lectures, STAT 234), <i>Harvard University</i>	<i>Apr 2022</i>
	L2. Offline off-policy reinforcement learning (STAT 234) <i>Harvard University</i> .	<i>Feb 2022</i>
	L3. Revisiting complexity and the bias-variance tradeoff (STAT 212) <i>UC Berkeley</i> .	<i>Apr 2021</i>
	L4. Introduction to ensemble methods in machine learning (EECS 189), <i>UC Berkeley</i> .	<i>Oct 2019</i>
	L5. Introduction to boosting methods (STAT 154), <i>UC Berkeley</i> .	<i>Apr 2019</i>
ACADEMIC SERVICES	<b>Undergraduate Research Mentoring</b>	
	• UC Berkeley, One student that led to a co-authored journal publication	<i>2020–2021</i>
	• Harvard, Two students with three co-authored submissions in preparation	<i>2022–</i>
	<b>Institutional Mentoring</b>	
	• MIT Institute for Data, Systems, & Society (IDSS) Postdoc Mentors for two <i>PhD</i> students	<i>2022–</i>
	• UC Berkeley Artificial Intelligence Research (BAIR) Buddies for two <i>incoming PhD</i> students	<i>2020–2021</i>
	• UC Berkeley BAIR Mentoring Program for five <i>undergraduates</i>	<i>2017–2021</i>
	• IIT Bombay Student Mentoring Program (ISMP) for twelve <i>incoming undergraduates</i>	<i>2013–2014</i>
	• IIT Bombay Academic Mentoring Program (DAMP) for four <i>sophomores &amp; juniors</i>	<i>2012–2014</i>
	• IIT Bombay Intensive Mentoring Program for thirty <i>undergraduates</i>	<i>2012–2013</i>
	<b>Committees</b>	
	• Member, Committee on Equality and Diversity, IMS	<i>2022–</i>
	<b>Scientific Meetings</b>	
	• Chair, New Researchers Group Session, IMS Annual Meeting	<i>2022</i>
	• Chair, Statistical Machine Learning Session, IMS Annual Meeting	<i>2022</i>
	• Mentor, Summer Institute on Just-in-Time Adaptive Interventions via MRTs	<i>2021</i>
	<b>Graduate Admissions</b>	
	• EECS Graduate Admissions Committee, MIT	<i>2021</i>
	• EECS Graduate Admissions Committee, UC Berkeley	<i>2018–2020</i>
	<b>Reviewing Activities</b>	
	• <i>Journals</i> : JMLR, IEEE-IT, JRSSB, Bernoulli, HDSR, Stats & Comp., SIAM, MOR, Jour. of Causal Inference	
	• <i>Conferences</i> : COLT, ICML, NeurIPS, AISTATS, FOCS, STOC, SODA, AAAI	
WORK EXPERIENCE	<b>Microsoft Research</b> , Research Intern with Lester Mackey, New England, USA	<i>2019</i>
	Mist Systems, Juniper Networks, Data Science Intern, Cupertino, USA	<i>2017</i>
	WorldQuant Research, Senior Quantitative Researcher, Mumbai, India	<i>2014–2015</i>
	<b>Stanford University</b> , Research Intern with Prof. Balaji Prabhakar, USA	<i>2013</i>
	Ivy Mobility, Data Science Intern, Chennai, India	<i>2012</i>

## REFERENCES

SUSAN MURPHY

Professor, CS & Statistics

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DEVAVRAT SHAH

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LESTER MACKEY

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Adjunct Professor, Stanford University

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MARTIN WAINWRIGHT

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BIN YU

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