Rajat Sen

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Interests

Online Learning, Causal Inference, Interpretability, Scheduling Systems

EDUCATION

The University of Texas at Austin

Ph.D. in Electrical Engineering(Com-Net-Sys)

August. 2013 - Present

Adviser : Prof. Sanjay Shakkottai
GPA : Major - 4.00, Overall - 3.96

Indian Institute of Technology, Kharagpur

B. Tech in Electrical Engineering

2009 - 2013

• Cumulative GPA: **9.24/10**

• Graduated ranked 5^{th} out of 90 students.

Honors and Awards

- Jagadish Bose National Science Talent Search (JBNSTS) senior scholar.
- INLAKS IIT Kharagpur Scholar, 2011: Only 3 students were selected from all 6 IITs.
- Secured All India Rank **251** in IIT joint entrance examination 2009(out of about 3,50,000 candidates).
- Received certificate of distinction in Indian National Mathematics Olympiad (INMO 2009).
- DAAD WISE Scholar, 2012.
- Awarded Certificate of merit in National Standard Examination in Physics (NSEP) 2008-2009 for being in the Top 1 % out of 34707 candidates appearing.

Publications

- Sen, R., Suresh, A.T., Shanmugam, K., Dimakis, A. G., & Shakkottai, S. (2017). Model-Powered Conditional Independence Test. arXiv preprint arXiv:1709.06138. (Preprint - Accepted for publication in NIPS, 2017)
- Sen, R., Shanmugam, K., Dimakis, A.G. & Shakkottai, S.. (2017). Identifying Best Interventions through Online Importance Sampling. Proceedings of the 34th International Conference on Machine Learning (ICML, 2017)
- Sen, R., Shanmugam, K., Kocaoglu, M., Dimakis, A. & Shakkottai, S.. (2017). Contextual Bandits with Latent Confounders: An NMF Approach. Proceedings of the 20th International Conference on Artificial Intelligence and Statistics (AISTATS, 2017)
- Krishnasamy, S., Sen, R., Johari, R., & Shakkottai, S. (2016). Regret of Queueing Bandits. In Advances in Neural Information Processing Systems (NIPS, 2016) (pp. 1669-1677).
- Krishnasamy, S., Sen, R., Shakkottai, S., & Oh, S. (2016). Detecting Sponsored Recommendations. ACM Transactions on Modeling and Performance Evaluation of Computing Systems (TOMPECS), 2(1), 6. (A shorter version appeared as a poster paper in ACM Sigmetrics, 2015.)
- Khalid, A., Sen, R., & Chattopadhyay, A. (2013, July). Si-dfa: Sub-expression integrated deterministic finite automata for deep packet inspection. In High Performance Switching and Routing (HPSR), 2013 IEEE 14th International Conference on (pp. 164-170). IEEE.

PROJECTS AND INTERNSHIPS

Data Science Intern

(May 2016 - August 2016)

Amplero, Seattle, Washington.

- Worked on bandit optimization in a marketing platform that delivers targeted messages and offers to millions of client customers with advanced segmentation.
- Proposed algorithmic improvements to the machine learning platform that combines specialized decision trees and Thompson sampling. The decision trees are reconditioned from optimization data using techniques from causality.
- Proposed novel variance prediction techniques for samples generated from bagged trees.
- Created a statistical test suite for verifying the correctness and predicting future performance.

Model-Powered Conditional Independence Test

(Feb 2017 - Present)

WNCG, UT Austin

- Created a robust Conditional Independence tester by reducing the problem to binary classification
- Contains novel analysis of classification generalization bounds in the presence of near-i.i.d samples
- Python Package: https://github.com/rajatsen91/CCIT

Identifying Best Interventions through Online Importance Sampling (Dec 2016 - Feb 2017)

WNCG, UT Austin.

- Designed a successive reject strategy for best soft intervention identification (using importance sampling to capture information leakage among the interventions), with applications in online ad placement and biological gene regulatory networks.
- Performed empirical validation on the Flow-cytometry dataset and also used it for interpretability
 of Inception v3 deep network (Python code-base available on request).

Contextual Bandits with Latent Confounders

(May 2016 - Oct 2016)

WNCG, UT Austin.

- Proposed a contextual bandit model with latent confounding contexts lying in lower dimension.
- Designed a non-negative matrix factorization based bandit algorithm with close to optimal regret guarantees. These are the first regret guarantees for matrix factorization with bandit feedback with rank greater than one.
- Implemented the algorithm in python and showed superior performance on MovieLens and Book Crossing data-sets when compared to baseline algorithms.

Bandit Algorithms for Queuing Systems

(May 2015 - May 2016)

WNCG, UT Austin.

- Developed scheduling algorithms for a stochastic switch network with semi-bandit feedback which gives first of its kind finite time queue length guarantees, with applications in wireless scheduling and crowd sourcing.
- We provide matching lower bounds and simulate results on switch networks.

Detecting Sponsored Recommendations

(November 2014 - August 2015)

WNCG, UT Austin.

- Designed an algorithm to detect bias of a recommendation engine towards certain sponsored items disguised as recommendations.
- Developed a model to describe a wide range of recommendation systems.
- Provided theoretical guarantees on the performance of the algorithm against various parameters like the number of users, the fraction of biased recommendations and the number of items.
- Simulated the performance of the algorithm on movielens, netflix and amazon datasets, with encouraging results.
- Github Link: https://github.com/rajatsen91/recosys_matlab

TEACHING EXPERIENCE **Teaching Assistant**, EE $351\mathrm{K}$ - Probability and Stochastic Processes University of Texas at Austin

Fall 2013

Relevant Coursework

- Graduate Courses: Probability and Stochastic processes, Randomized algorithms, Combinatorics and Graph Theory, Data Mining and Predictive Learning, Markov Chains and Mixing times, Large-scale Optimization, Theory of Probability, Advanced Algorithms, Information Theory
- MOOC: Big Data using Spark(edX), Scalable Machine Learning using Spark(edX)

TECHNICAL SKILLS

- Programming Tools: Python (Sklearn, Pandas, Numpy, PySpark), PyTorch, C, C++, MAT-LAB
- Documentation: LaTeX, MS Word