Publications

• Scalable Greedy Support Selection via Weak Submodularity. Rajiv Khanna, Ethan Elenberg, Alex Dimakis, Sahand Neghaban, Joydeep Ghosh. AISTATS 2017 (to appear)

rajivak@utexas.edu

Ph: 5129476729

- Information Projection and Approximate Inference for Structured Sparse Variables. Rajiv Khanna, Joydeep Ghosh, Russell Poldrack, Oluwasanmi Koyejo AISTATS 2017 (to appear)
- A Unified Analysis of Frank Wolfe and Matching Pursuit. Francesco Locatello, Rajiv Khanna, Michael Tschannen, Martin Jaggi AISTATS 2017 (to appear)
- Pursuits in Structured Non-Convex Matrix Factorizations. Rajiv Khanna, Francesco Locatello, Michael Tschannen, Martin Jaggi. Arxiv Report.
- A Deflation Method for Structured Probabilistic PCA. Rajiv Khanna, Joydeep Ghosh, Russell A. Poldrack, Oluwasanmi Koyejo **SDM 2017** (to appear)
- Examples are not Enough, Learn to Criticize! Criticism for Interpretability. Been Kim*, Rajiv* Khanna, Oluwasanmi Koyejo*. NIPS 2016 (Oral)
- Restricted Strong Convexity implies Weak Submodularity. Ethan Elenberg, Rajiv Khanna, Alex Dimakis, Sahand Neghaban. NIPS 2016 Workshop on Learning in High Dimensions with Structure.
- Towards a Better Understanding of Predict and Count Models. S. Sathiya Keerthi, Tobias Schnabel, Rajiv Khanna. Arxiv report.
- Sparse Submodular Probabilistic PCA. Rajiv Khanna, Joydeep Ghosh, Russell A. Poldrack, Oluwasanmi O. Koyejo AISTATS 2015 (Oral)
- A Deflation Method for Probabilistic PCA. Rajiv Khanna, Joydeep Ghosh, Russell A. Poldrack, Oluwasanmi Koyejo. NIPS 2015 Workshop on Advances in Approximate Bayesian Inference.
- On Prior Distributions and Approximate Inference for Structured Variables. Oluwasanmi O. Koyejo, Rajiv Khanna, Joydeep Ghosh, Russell A. Poldrack NIPS 2014
- Parallel Matrix Factorization for Binary Response. Rajiv Khanna, Deepak Agarwal, Liang Zhang and Beechung Chen. IEEE BigData 2013
- Estimating Rates of Rare Events with Multiple Hierarchies through Scalable Loglinear Models. Deepak Agarwal*; Rahul Agrawal*; Rajiv Khanna*; Nagaraj Kota*. **KDD 2010**
- Translating Relevance Scores to Probabilities for Contextual Advertising. Deepak Agarwal*; Evgeniy Gabrilovich*; Rob Hall*; Vanja Josifovski*; Rajiv Khanna*. CIKM 2009
- Structured Learning for Non-Smooth Ranking Losses. Soumen Chakrabarti, Rajiv Khanna, Uma Sawant, Chiru Bhattacharyya. **KDD 2008**

Professional Experience

Program Committee/ Reviewer : NIPS 2016, WWW 2017, Workshop on Advances in Approx. Bayesian Inference 2015/2016.

ETH Zurich (Summer 2015)

• Generalized Pursuit algorithms. (Mentor: Martin Jaggi)

Microsoft Research (Summer 2014)

• Word Embeddings: Worked on a text mining problem that involves representation of words as vectors which can then be used in various ways such as features for a prediction problem.

LinkedIn Inc. (Summer 2013)

 Online features for contextual advertising – Explored inclusion of online features within limitations of the existing Ad serving infrastructure. Proposals showed lifts, slated for bucket testing.

Research Engineer at Yahoo! Labs Bangalore (July 2008-July 2012)

- Web scale recommendation systems Part of team that developed and tested matrix factorization based recommendation systems on real world data of the order of ~1 billion events using map-reduce. The models were tested over real- world data and gave significant lifts in Click-Through Rates (CTR) when compared to the other state-of-the-art methods.
 - Modeling skewed data Implemented models that address the challenges of modeling CTR and CTR-derived goals. These problems are difficult because of the scale, low CTR, non-trivial derived goals and object hierarchies.
 - <u>Information corroboration</u> Given data extracted from different web sources having potentially conflicting information, we developed graph-transduction algorithms to filter the "most likely correct" info. The methods we developed passed production-level quality tests and as a part of an extraction pipeline enriched content information for entities in the order of hundreds of thousands.

IIT Bombay

- <u>Learning to rank for non-smooth ranking losses</u> (MTech thesis; Guide: Prof Soumen Chakrabarti): Our objective is to learn a real-valued ranking function from the given labeled training data. Using structured learning paradigm, we directly optimize for non-smooth ranking losses like Mean Reciprocal Rank (MRR) and Normalized Discounted Cumulative Gain (NDCG) etc.
- Uncertainty in databases (MTech Seminar)

Relevant Course Work

- (*UT Austin*) Probability and Stochastic Processes, Large Scale Learning, Statistical Modeling II, Optimization in Engg Systems, Game Theory, Measure theoretic probability, Large Scale Optimization (Audit), Randomized Algorithms (Audit),
- (*IIT Bombay*) Statistical Foundations of Machine Learning, Web Mining and Extraction, Artificial Intelligence, Probabilistic Graphical Models, Algorithms & Complexity, Object Oriented Programming, Convex Optimization(Audit).

Technical Skills

• R, Matlab, Perl, Python, Shell scripting, C, Java, Hadoop, Pig.

Academics

Degree	University	Year	Grade
PhD	UT Austin	2012-	4.0/4.0 (GPA)
M.Tech (Masters)	IIT Bombay	2006-2008	9.78/10 (CPI)
B.Tech (undergrad)	NIT Jallandhar	2002-2006	82.34/100 (%)

Achievements/Awards:

- Travel Award (NIPS 2014)
- One of the two students (out of 85) awarded the Phillips Scholarship, on basis of academic performance at Department of CSE, IIT Bombay.
- Rank 1 amongst 55 CS MTech students at IIT Bombay throughout the course, awarded "most outstanding student" of outgoing 2008 Masters batch. Top-5 in undergrad.
- Ranked 97 in GATE (CS) in 2006 out of over 22000 students
- Ranked 452 (stage 1) and 4292 (stage 2) in IIT-JEE undergrad entrance 2002 (forgone admission in favor of studying computer science).
- Test scores GRE (Nov 2012) 330/340, TOEFL (Nov 2012) 117/120