Dong Hu

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INTERESTS

Matrix Completion, Sketching, Machine Learning, High-dimensional Statistics, Applied Linear Algebra, Randomized Algorithms.

EDUCATION

Rensselaer Polytechnic Institute

Ph.D. in Computer Science

GPA: 3.9/4.0

Summer 2019 – Spring 2024 (expected)

Advisor: Prof. Alex Gittens

Rensselaer Polytechnic Institute Bachelor of Science in Mathematics Bachelor of Science in Computer Science GPA: 3.86/4.0, Dean's Honor List

RESEARCH EXPERIENCE

Rensselaer Polytechnic Institute

Summer 2019-present

Spring 2016 – Spring 2019 Advisor: *Prof.* Jeffery Banks

Advisor: Prof. Heng Ji

Graduate Research Assistant, Computer Science Department

- Proposed a proximal regularized sketched alternating least squares (Tucker-ALS) algorithm for low rank Tucker decomposition of large tensors, and prove that a sublinear rate of convergence of proximally regularized sketched CPD algorithms also holds for the proposed algorithm. Showed that the iterative nature of the Tucker-ALS approach can be exploited algorithmically to choose more performant sketching rates at different iterations to improve the convergence of the overall algorithm.
- Proposed and investigated in a regression-based matrix completion algorithm (noisyCUR) for low budget matrix completion setting and experimentally verify the performance of our algorithm on both synthetic and real data, compared our algorithm with state-of-the-art Matrix completion algorithms. Paper accepted by ECML 2020.

IBM, Yorktown Heights

Summer 2020-present

 $Artificial\ Intelligence\ Research\ Collaboration(AIRC)\ scholar$

• Investigated sketching matrices that are obtained from bipartite graphs that are sparse, and explored two popular classes of sparse graphs, expander graphs and magical graphs. Proved that for a subspace with dimention k, the magical graph with left degree s=2 yields a $(1 \pm \varepsilon)$ ℓ_2 -subspace embedding if the number of right vertices (the sketch size) $m=\mathcal{O}(k^2/\varepsilon^2)$, and the expander graph with $s=\mathcal{O}(\log k/\varepsilon)$ yields a subspace embedding for $m=\mathcal{O}(k\log k/\varepsilon^2)$.

PUBLICATIONS D. Hu, S. Ubaru, A. Gittens, K. Clarkson, L. Horesh, and V. Kalantzis. "Sparse graph based sketching for fast numerical linear algebra." in *International Conference on Acoustics, Speech, and Signal Processing (ICASSP)*, 2021.

D. Hu, A. Gittens, and M. Magdon-Ismail, "NoisyCUR: An algorithm for two-cost budgeted matrix completion," in *Machine Learning and Knowledge Discovery in Databases - European Conference(ECML-PKDD)*, 2020