

CSSI Element: <A high performance suite of SVD related solvers for machine learning>

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1. SVD methods in distance-based learning

Motivation

- The need for investigation of the interplay between SVD solvers and ML algorithm
- Clarify the best practice guidelines
- Propose new algorithms /improve the current existing SVD related algorithms

Our contribution

We examine 3 categories

- 1. Kernel ridge regression and speedup methods
- 2. Matrix completion / Nuclear norm
- 3. Spectral clustering

Key research questions:

- 1. With different types of eigenvalue distribution, how to identify the termination conditions to make better ML performance?
- 2. What are the bottleneck in real dataset using SVD and how to speed up the algorithms?

2. Low-Rank Stopping Criteria for Block Parallel SVD

- Motivation
- Most SVD algorithms only provide a residual stopping criterion: limit the number of total iterations, matrix vector multiplications, or its execution time
- Lack guarantee on the accuracy

Our contribution

- 1. New stopping criteria for block parallel SVD algorithms
- 2. Provide heuristics for dynamically changing both block and restart sizes when necessary
- Show the performance in both synthetic and real-world applications

