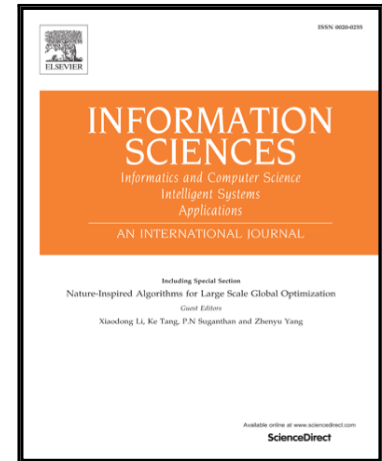


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Collaborative Filtering Using Multiple Binary Maximum Margin Matrix Factorizations

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Highlights

- In Maximum Margin Matrix Factorization(MMMF), ratings matrix with multiple discrete values is treated by specially extending hinge loss function to suit multiple levels. We view this process as analogous to extending two-class classifier to a unified multi-class classifier.
- Alternatively, multi-class classifier can be built by arranging multiple two-class classifiers in a hierarchical manner. In this paper, we investigate this aspect for collaborative filtering and propose an efficient and novel framework of multiple bi-level MMMFs.
- We propose an stage-wise matrix completion technique that makes use several bi-level MMMFs in a hierarchical fashion.
- We compare our method with nine well-known algorithms on two benchmark datasets and show that our method outperforms these methods on NMAE measure.
- We also show that our method yields latent factors of lower ranks and the trade-off between empirical and generalization error is low.

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