Convergence Analysis on Orthogonal Low Rank Approximation of Tensors

Problem setting

Determine unit vectors every u_r and λ_r to minimize

$$\left\| T - \sum_{r=1}^{R} \lambda_r \mathbf{u}_r^{(1)} \circ \cdots \circ \mathbf{u}_r^{(k)} \right\|_F^2.$$

Polar decomposition is applied for last μ factors

for
$$r = 1, 2, ..., R$$
, do
 $\mathbf{v}_{r,[p+1]}^{(\ell)} = T \circledast_{\ell} \left(\bigotimes_{i=1}^{\ell-1} \mathbf{u}_{r,[p+1]}^{(i)} \circ \bigotimes_{i=\ell+1}^{k} \mathbf{u}_{r,[p]}^{(i)} \right)$
 $\hat{\lambda}_{r,[p+1]}^{(\ell)} := \langle \mathbf{v}_{r,[p+1]}^{(\ell)}, \mathbf{u}_{r,[p]}^{(\ell)} \rangle$

- Best rank-R (R > 1) approximation for high-order tensors may not exist
- Consider orthogonal low rank approximation
- Including completely orthogonal and semi-orthogonal low rank

end for $[U_{[p+1]}^{(\ell)}, S_{[p+1]}^{(\ell)}] = \text{poldec}(V_{[p+1]}^{(\ell)} \wedge_{[p+1]}^{(\ell)})$

Convergence

- the generalized Rayleigh quotients are bounded and monotone increasing iterates globally converge to local minimum

approximation

Numerical result

Algorithm

- SVD is involved for the first $k - \mu$ factors
- for r = 1, 2, ..., R, $C_{r,[p+1]}^{(\ell)} = T \circledast_{\beta_{\ell}} \left(\bigotimes_{i=1}^{\ell-1} \mathbf{u}_{r,[p+1]}^{(i)} \circ \bigotimes_{i=\ell+2}^{k} \mathbf{u}_{r,[p]}^{(i)} \right)$ $[\mathbf{u}, s, \mathbf{v}] = \operatorname{svds}(C_{r,[p+1]}^{(\ell)}, 1)$ end for





[1] J. CHEN AND Y. SAAD, On the tensor SVD and the optimal

low rank orthogonal approximation of tensors, SIAM J. Matrix Anal. Appl., 30 (2008/09), pp. 1709–1734.

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