

Fixed points theorems for nonnegative tensors and Newton method

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Overview

- 1 Perron-Frobenius theorem for *irreducible* nonnegative tensors.

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- 2 Diagonal scaling of nonnegative tensors to tensors with given rows, columns and depth sums.

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$\pm\sigma_i(A)$, $i = 1, \dots$ are critical values of $f(\mathbf{x}, \mathbf{y}) = \mathbf{x}^T A\mathbf{y}$

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Proof: $A^\top A, AA^\top$ are irreducible

Rank one approximations for 3-tensors

$$\mathbb{R}^{m \times n \times l} \text{ IPS: } \langle \mathcal{A}, \mathcal{B} \rangle = \sum_{i,j,k} a_{i,j,k} b_{i,j,k}, \quad \|\mathcal{T}\|_2 = \sqrt{\langle \mathcal{T}, \mathcal{T} \rangle}$$
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X subspace of $\mathbb{R}^{m \times n \times l}$, $\mathcal{X}_1, \dots, \mathcal{X}_d$ an orthonormal basis of **X**

$$\mathbf{P}_X(\mathcal{T}) = \sum_{i=1}^d \langle \mathcal{T}, \mathcal{X}_i \rangle \mathcal{X}_i, \quad \|\mathbf{P}_X(\mathcal{T})\|_2^2 = \sum_{i=1}^d \langle \mathcal{T}, \mathcal{X}_i \rangle^2$$

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How many distinct singular values are for a generic tensor?

ℓ_p maximal problem and Perron-Frobenius

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$$\|(x_1, \dots, x_n)^\top\|_p := (\sum_{i=1}^n |x_i|^p)^{\frac{1}{p}}$$

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$\mathcal{T} \times \mathbf{x} \otimes \mathbf{z} = \lambda \mathbf{y}^{p-1}$, $\mathcal{T} \times \mathbf{x} \otimes \mathbf{y} = \lambda \mathbf{z}^{p-1}$ ($p = \frac{2t}{2s-1}$, $t, s \in \mathbb{N}$)

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See L.-H. Lim 2005 for more general results

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Assume that $\mathcal{T} \geq 0$. Then $\mathbf{x}, \mathbf{y}, \mathbf{z} \geq 0$

For which values of p we have an analog of Perron-Frobenius theorem?, **UNIQUENESS**

Yes, for $p \geq 3$, No, for $p < 3$,
Friedland-Gauber-Han [5]

Outline of the proof

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Define: $F : \mathbb{R}_+^m \times \mathbb{R}_+^n \times \mathbb{R}_+^l \rightarrow \mathbb{R}_+^m \times \mathbb{R}_+^n \times \mathbb{R}_+^l$:

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If F completely irreducible, i.e. F^N maps nonzero nonnegative vectors to positive, nonnegative eigenvector is unique and positive

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$PAQ = \begin{bmatrix} A_{11} & 0 \\ A_{21} & A_{22} \end{bmatrix}$ then the columns sums of \mathbf{c} corresponding to the columns of A_{11} are strictly less than the row sums of \mathbf{r} of the rows of A_{11} .

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$\sum_{i=1}^m r_i = \sum_{j=1}^n c_j = \sum_{k=1}^l d_k$

Find nec. and suf. conditions for scaling:

$\mathcal{T}' = [t_{i,j,k} e^{x_i+y_j+z_k}]$, $\mathbf{x}, \mathbf{y}, \mathbf{z}$ such that \mathcal{T}' has given row, column and depth sum

Solution: Convert to the minimal problem:

$\min_{\mathbf{r}^\top \mathbf{x} = \mathbf{c}^\top \mathbf{y} = \mathbf{d}^\top \mathbf{z} = 0} f_{\mathcal{T}}(\mathbf{x}, \mathbf{y}, \mathbf{z})$, $f_{\mathcal{T}}(\mathbf{x}, \mathbf{y}, \mathbf{z}) = \sum_{i,j,k} t_{i,j,k} e^{x_i+y_j+z_k}$

Any critical point of $f_{\mathcal{T}}$ on $\mathcal{S} := \{\mathbf{r}^\top \mathbf{x} = \mathbf{c}^\top \mathbf{y} = \mathbf{d}^\top \mathbf{z} = 0\}$ gives rise to a solution of the scaling problem (Lagrange multipliers)

$f_{\mathcal{T}}$ is convex

$f_{\mathcal{T}}$ is strictly convex implies \mathcal{T} is not decomposable: $\mathcal{T} \neq \mathcal{T}_1 \oplus \mathcal{T}_2$.

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Are variants Brualdi theorem hold in the tensor case?

Rescaling versus Newton method

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Hence Newton method has a quadratic convergence versus linear convergence of Sinkhorn algorithm

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True for matrices too

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