

# Power Method for the maximal $\ell^{p_1, \dots, p_m}$ -singular value of non-negative tensor

## Tensor projective norm

Given a tensor  $f \in \mathbb{R}^{d_1 \times d_2 \times \dots \times d_m}$  of order  $m$  and  $1 < p_1, \dots, p_m < \infty$ , we propose in our paper, a power method type algorithm for computing a maximizer of the following problem:

$$\|f\|_{p_1, \dots, p_m} := \max_{\mathbf{x}_1, \dots, \mathbf{x}_m \neq 0} \frac{|f(\mathbf{x}_1, \dots, \mathbf{x}_m)|}{\|\mathbf{x}_1\|_{p_1} \cdot \dots \cdot \|\mathbf{x}_m\|_{p_m}}$$

where  $\|\mathbf{z}\|_{p_i} := \left(\sum_{k=1}^{n_i} |z_k|^{p_i}\right)^{1/p_i}$  denotes the usual  $p_i$ -norm on  $\mathbb{R}^{n_i}$  and  $f(\mathbf{x}_1, \dots, \mathbf{x}_m)$  is the multi-linear form associated to the tensor  $f$ , i.e.

$$f(\mathbf{x}_1, \dots, \mathbf{x}_m) := \sum_{j_1 \in [d_1], \dots, j_m \in [d_m]} f_{j_1, \dots, j_m} x_{1, j_1} \cdot \dots \cdot x_{m, j_m} \quad \forall (\mathbf{x}_1, \dots, \mathbf{x}_m) \in \mathbb{R}^{d_1} \times \mathbb{R}^{d_2} \times \dots \times \mathbb{R}^{d_m}$$

with  $[d_i] := \{1, \dots, d_i\}$ .

**Matrix case:** When the tensor is of order  $m = 2$ , i.e.  $f = A \in \mathbb{R}^{d_1 \times d_2}$  is a matrix, the maximization problem above reduces to the so-called matrix  $p, q$ -norm, namely

$$\|A\|_{p, q} := \max_{\mathbf{y}, \mathbf{z} \neq 0} \frac{|\mathbf{y}^t A \mathbf{z}|}{\|\mathbf{y}\|_p \|\mathbf{z}\|_q} = \max_{\mathbf{z} \neq 0} \frac{\|A \mathbf{z}\|_{p'}}{\|\mathbf{z}\|_q} = \max_{\mathbf{y} \neq 0} \frac{\|A^t \mathbf{y}\|_{q'}}{\|\mathbf{y}\|_p}$$

with  $1 < p, q < \infty$ .

**Remark.** Provided that the convergence assumptions are fulfilled, the algorithm produces two monotonic sequences  $(\lambda_-^k), (\lambda_+^k) \subset \mathbb{R}$  such that

$$\lambda_-^k \leq \lambda_-^{k+1} \leq \|f\|_{p_1, \dots, p_m} \leq \lambda_+^{k+1} \leq \lambda_+^k \quad \forall k = 1, 2, \dots$$

and thus can also be used to simply obtain an upper bound on  $\|f\|_{p_1, \dots, p_m}$ . We use these sequences as stopping criteria, i.e. the algorithm runs until  $\frac{\lambda_+^k - \lambda_-^k}{\lambda_+^k} < \text{tol}$ . For an explicit expression of  $\lambda_-^k$  and  $\lambda_+^k$  we refer to the paper cited below.

## Convergence guarantee

The method is guaranteed to converge to a global maximizer of the optimization problem above if:

- $f$  is non-negative, i.e.  $f_{j_1, \dots, j_m} \geq 0$  for all  $j_1 \in [d_1], \dots, j_m \in [d_m]$ .
- $1 < p_1, \dots, p_m < \infty$  and there exists  $i \in [m]$  such that

$$m - 1 \leq (p'_i - 1) \left( \min_{k \in [m] \setminus \{i\}} p_k - (m - 1) \right).$$

*Example:*

- $p_1 = 1 + t, p_2 = \dots = p_m \geq (m-1)(1 + 1/t), t > 0.$
- $p_1 = \dots = p_{m-1} = (m-1) + s, p_m \geq \frac{m-1+s}{s}, s > 0.$

c) The tensor  $f$  is **weakly irreducible**. That is, the undirected  $m$ -partite graph  $G(T) = (V, E)$  is connected where

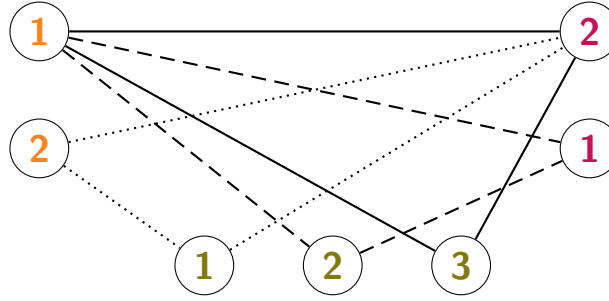
$$V := [d_1] \sqcup [d_2] \sqcup \dots \sqcup [d_m],$$

is the disjoint union of  $[d_1], [d_2], \dots, [d_m]$  and

$$(j_k, j_l) \in E \iff \forall i \in [d] \setminus \{k, l\}, \exists j_i \in [d_i] \text{ such that } f_{j_1, \dots, j_d} > 0.$$

*Example:*

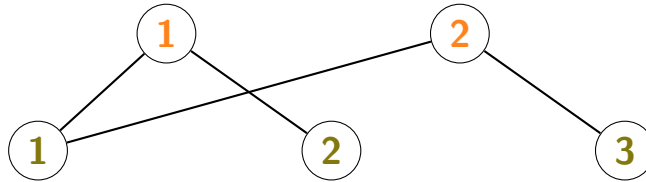
$$f \in \mathbb{R}^{2 \times 3 \times 2} \quad \text{with} \quad \underline{f_{1,2,1}} > 0 \quad \underline{f_{2,1,2}} > 0 \quad \underline{f_{1,3,2}} > 0 \quad \text{and} \quad f_{i,j,k} = 0 \text{ else.}$$



**Matrix case:** When the tensor  $f = A \in \mathbb{R}^{d_1 \times d_2}$  is a matrix, then the three conditions above reduce to:

- $A$  is non-negative, i.e.  $A_{i,j} \geq 0$  for every  $i \in [d_1], j \in [d_2]$ .
- $1 < p, q < \infty$  and  $(p-1)(q-1) \geq 1$ .
- $A$  is **weakly irreducible** (see definition above).

*Example:* If  $A = \begin{pmatrix} 1 & 2 & 0 \\ \pi & 0 & 4 \end{pmatrix} \in \mathbb{R}^{2 \times 3}$ , then the undirected graph  $G(A)$  is given by



and therefore  $A$  is weakly irreducible because  $G(A)$  is connected.

## Download

Please include a reference to our preprint

- A. Gautier and M. Hein  
Tensor norm and maximal singular vectors of non-negative tensors - a Perron-Frobenius theorem, a Collatz-Wielandt characterization and a generalized power method  
arXiv:1503.01273

if you find this code useful for your research.