

Rank-one Boolean tensor factorization and the multilinear polytope

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Boolean tensor factorization: introduction

- A tensor of **order** N is an N -dimensional array. Factorizations of high-order tensors, i.e., $N \geq 3$, as products of low-rank matrices, have applications in signal processing, numerical linear algebra, computer vision, data mining, neuroscience, and elsewhere.
- We consider the problem of factorizing a high-order tensor with **binary** entries, referred to as a **binary tensor**.
- In **Boolean tensor factorization (BTF)**, the binary tensor is approximated by products of **low rank binary matrices** using **Boolean algebra**.
- Applications include neuro-imaging, recommendation systems, topic modeling, and sensor network localization
- BTF is **NP-hard** in general; all existing methods rely on heuristics without any guarantee on the quality of the solution.

Boolean tensor factorization: problem statement

- For simplicity, we focus on tensors of **order three**.
- The **(Boolean) rank** of a binary tensor \mathcal{G} is the smallest integer r such that there exist $3r$ binary vectors x^p, y^p, z^p , for $p \in [r]$, with

$$\mathcal{G} = \bigvee_{p=1}^r (x^p \otimes y^p \otimes z^p),$$

where \vee denotes the component-wise “or” operation, and \otimes denotes the vector outer product.

- The **rank- r BTF** is the problem of finding the closest rank- r binary tensor to a binary tensor. Given a $n \times m \times l$ binary tensor \mathcal{G} and an integer r , find $3r$ binary vectors $x^p \in \{0, 1\}^n$, $y^p \in \{0, 1\}^m$, $z^p \in \{0, 1\}^l$, for all $p \in [r]$, that minimize

$$\left\| \mathcal{G} - \bigvee_{p=1}^r x^p \otimes y^p \otimes z^p \right\|^2,$$

where the Frobenius norm of \mathcal{G} is defined as $\|\mathcal{G}\| := \sqrt{\sum_{i,j,k} w_{ijk}^2}$.

Rank-one Boolean tensor factorization

- **Rank-one BTF**: the simplest case of BTF with $r = 1$:

$$\begin{aligned} \min \quad & \left\| \mathcal{G} - x \otimes y \otimes z \right\|^2 \\ \text{s.t.} \quad & x \in \{0, 1\}^n, \quad y \in \{0, 1\}^m, \quad z \in \{0, 1\}^l. \end{aligned}$$

- Rank-one BTF is **NP-hard** and no algorithm with theoretical guarantees is known for this problem.
- Define $S_0 := \{(i, j, k) \in [n] \times [m] \times [l] : g_{ijk} = 0\}$, $S_1 := \{(i, j, k) \in [n] \times [m] \times [l] : g_{ijk} = 1\}$, and $w_{ijk} := x_i y_j z_k$, for $i \in [n]$, $j \in [m]$, $k \in [l]$.
- Rank-one BTF can be written, in an extended space, as the problem of minimizing a linear function over a **highly structured multilinear set**:

$$\begin{aligned} \min \quad & \sum_{(i,j,k) \in S_0} w_{ijk} + \sum_{(i,j,k) \in S_1} (1 - w_{ijk}) \\ \text{s.t.} \quad & w_{ijk} = x_i y_j z_k, \quad \forall i \in [n], j \in [m], k \in [l] \\ & x \in \{0, 1\}^n, \quad y \in \{0, 1\}^m, \quad z \in \{0, 1\}^l. \end{aligned}$$

Multilinear sets and polytopes

- With any hypergraph $G = (V, E)$, we associate a Multilinear set \mathcal{S}_G defined as:

$$\mathcal{S}_G = \{(u, w) \in \{0, 1\}^{|V|} \times \{0, 1\}^{|E|} : w_e = \prod_{v \in e} u_v, e \in E\},$$

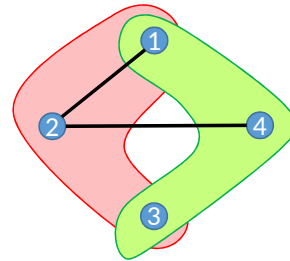
and the **Multilinear polytope** MP_G is the **convex hull** of \mathcal{S}_G .

$$w_{12} = u_1 u_2$$

$$w_{24} = u_2 u_4$$

$$w_{123} = u_1 u_2 u_3$$

$$w_{134} = u_1 u_3 u_4$$

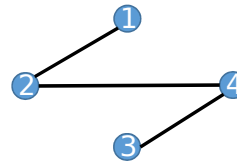


- For **quadratic sets**, we obtain the **graph representation** of the **Boolean quadric polytope** (Padberg, 89)

$$w_{12} = u_1 u_2$$

$$w_{24} = u_2 u_4$$

$$w_{34} = u_3 u_4$$



- There are interesting connections between the **complexity of MP_G** and the **acyclicity degree** of its hypergraph.

Standard linearization of Rank-one BTF

- A simple LP relaxation of Rank-one BTF can be obtained by replacing **each multilinear term** $w_{ijk} = x_i y_j z_k$, $x_i, y_j, z_k \in \{0, 1\}$, by its convex hull:

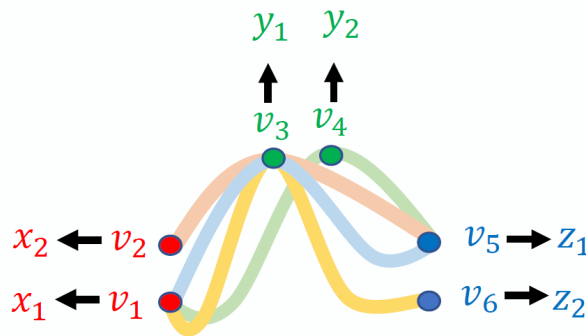
$$\min \sum_{(i,j,k) \in S_0} w_{ijk} + \sum_{(i,j,k) \in S_1} (1 - w_{ijk}) \quad (\text{sLP})$$

$$\text{s.t. } w_{ijk} \leq x_i, w_{ijk} \leq y_j, w_{ijk} \leq z_k, \quad \forall (i, j, k) \in S_1$$

$$w_{ijk} \geq 0, w_{ijk} \geq x_i + y_j + z_k - 2, \quad \forall (i, j, k) \in S_0$$

$$(x, y, z) \in [0, 1]^{n+m+l}.$$

- Stronger LP relaxations can be obtained by **convexifying multiple multilinear terms** at a time.



$$\tilde{G} = (\tilde{V}, \tilde{E})$$

$$\tilde{V} = \{v_1, v_2, v_3, v_4, v_5, v_6\}$$

$$\tilde{E} = \{e_0, e_1, e_2, e_3\}$$

Rank-one BTF and the multilinear polytope

- **Proposition:** The facet description of $\text{MP}_{\tilde{G}}$ is given by:

$$w_e \leq u_v, \quad \forall e \in \tilde{E}, \forall v \in e$$

$$w_e \geq 0, \quad \sum_{v \in e} u_v - w_e \leq 2, \quad \forall e \in \tilde{E}$$

$$u_v \leq 1, \quad \forall v \in \tilde{V}$$

$$u_{e_0 \setminus e} + w_e - w_{e_0} \leq 1, \quad \forall e \in \tilde{E} \setminus \{e_0\}$$

$$u_{e \setminus e_0} - w_e + w_{e_0} \leq 1, \quad \forall e \in \tilde{E} \setminus \{e_0\}$$

$$-u_{e \cap e'} + w_e + w_{e'} - w_{e_0} \leq 0, \quad \forall e \neq e' \in \tilde{E} \setminus \{e_0\}$$

$$\sum_{v \in \tilde{V}} u_v - \sum_{e \in \tilde{E} \setminus \{e_0\}} w_e + w_{e_0} \leq 4.$$

- **Proposition:** All inequalities defining facets of $\text{MP}_{\tilde{G}}$ are facet-defining for the multilinear polytope of rank-one BTF, once we associate u_{v_1}, u_{v_2} to any two x variables, u_{v_3}, u_{v_4} to any two y variables, and u_{v_5}, u_{v_6} to any two z variables

LP relaxations for Rank-one BTF

- A stronger LP relaxation of Rank-one BTF is given by

$$\begin{aligned}
 \min \quad & \sum_{(i,j,k) \in S_0} w_{ijk} + \sum_{(i,j,k) \in S_1} (1 - w_{ijk}) & (\text{cLP}) \\
 \text{s.t.} \quad & w_{ijk} \leq x_i, \quad w_{ijk} \leq y_j, \quad w_{ijk} \leq z_k, \quad \forall (i, j, k) \in S_1 \\
 & w_{ijk} \geq 0, \quad w_{ijk} \geq x_i + y_j + z_k - 2, \quad \forall (i, j, k) \in S_0 \\
 & w_{i'jk} - w_{ijk} \leq 1 - x_i, \quad \forall (i, j, k) \in S_0, (i', j, k) \in S_1 \\
 & w_{ij'k} - w_{ijk} \leq 1 - y_j, \quad \forall (i, j, k) \in S_0, (i, j', k) \in S_1 \\
 & w_{ijk'} - w_{ijk} \leq 1 - z_k, \quad \forall (i, j, k) \in S_0, (i, j, k') \in S_1 \\
 & w_{ij'k} + w_{ijk'} - w_{ijk} \leq x_i, \quad \forall (i, j, k) \in S_0, (i, j', k) \in S_1, (i, j, k') \in S_1 \\
 & w_{i'jk} + w_{ijk'} - w_{ijk} \leq y_j, \quad (i, j, k) \in S_0, \forall (i', j, k) \in S_1, (i, j, k') \in S_1 \\
 & w_{i'jk} + w_{ij'k} - w_{ijk} \leq z_k, \quad \forall (i, j, k) \in S_0, (i', j, k) \in S_1, (i, j', k) \in S_1 \\
 & x_i + x_{i'} + y_j + y_{j'} + z_k + z_{k'} + w_{ijk} - w_{i'jk} - w_{ij'k} - w_{ijk'} \leq 4, \\
 & \quad \quad \quad \forall (i', j, k) \in S_0, (i, j', k) \in S_0, (i, j, k') \in S_0, (i, j, k) \in S_1 \\
 & (x, y, z) \in [0, 1]^{n+m+l}.
 \end{aligned}$$

LP relaxations for Rank-one BTF

- We analyze the theoretical performance of the following relaxation of Problem (cLP):

$$\begin{aligned}
 \min \quad & \sum_{(i,j,k) \in S_0} w_{ijk} + \sum_{(i,j,k) \in S_1} (1 - w_{ijk}) && \text{(fLP)} \\
 \text{s.t.} \quad & w_{ijk} \leq x_i, \quad w_{ijk} \leq y_j, \quad w_{ijk} \leq z_k, && \forall (i, j, k) \in S_1 \\
 & w_{ijk} \geq 0, \quad w_{ijk} \geq x_i + y_j + z_k - 2, && \forall (i, j, k) \in S_0 \\
 & w_{i'jk} - w_{ijk} \leq 1 - x_i, && \forall (i, j, k) \in S_0, (i', j, k) \in S_1 \\
 & w_{ij'k} - w_{ijk} \leq 1 - y_j, && \forall (i, j, k) \in S_0, (i, j', k) \in S_1 \\
 & w_{ijk'} - w_{ijk} \leq 1 - z_k, && \forall (i, j, k) \in S_0, (i, j, k') \in S_1 \\
 & (x, y, z) \in [0, 1]^{n+m+l}.
 \end{aligned}$$

Recovery under random models

- Given an LP relaxation of Rank-one BTF and a random model for noise in the input tensor, **what is the maximum level of noise under which the relaxation recovers the ground truth with high probability?** (probability tending to 1 as $n, m, l \rightarrow \infty$.)
- **Random corruption model for rank-one BTF:** consider binary vectors $\bar{x} \in \{0, 1\}^n$, $\bar{y} \in \{0, 1\}^m$, $\bar{z} \in \{0, 1\}^l$ and define the **ground truth** rank-one tensor $\bar{W} = (w_{ijk}) := \bar{x} \otimes \bar{y} \otimes \bar{z}$. Given $p \in [0, 1]$, the noisy tensor \mathcal{G} is as follows: for each $(i, j, k) \in [n] \times [m] \times [l]$, g_{ijk} is corrupted with probability p , i.e., $g_{ijk} := 1 - \bar{x}_i \bar{y}_j \bar{z}_k$, and g_{ijk} is not corrupted with probability $1 - p$, i.e., $g_{ijk} := \bar{x}_i \bar{y}_j \bar{z}_k$. we focus on the case where p is a constant.
- Denote by $r_{\bar{x}}$ (resp. $r_{\bar{y}}$, $r_{\bar{z}}$) the ratio of ones in \bar{x} (resp. \bar{y} , \bar{z}) to the number of elements in \bar{x} (resp. \bar{y} , \bar{z}). Assume that $r_{\bar{x}}, r_{\bar{y}}, r_{\bar{z}}$ are positive constants and let $r_{\bar{w}} := r_{\bar{x}} r_{\bar{y}} r_{\bar{z}}$.

Information theoretic limits

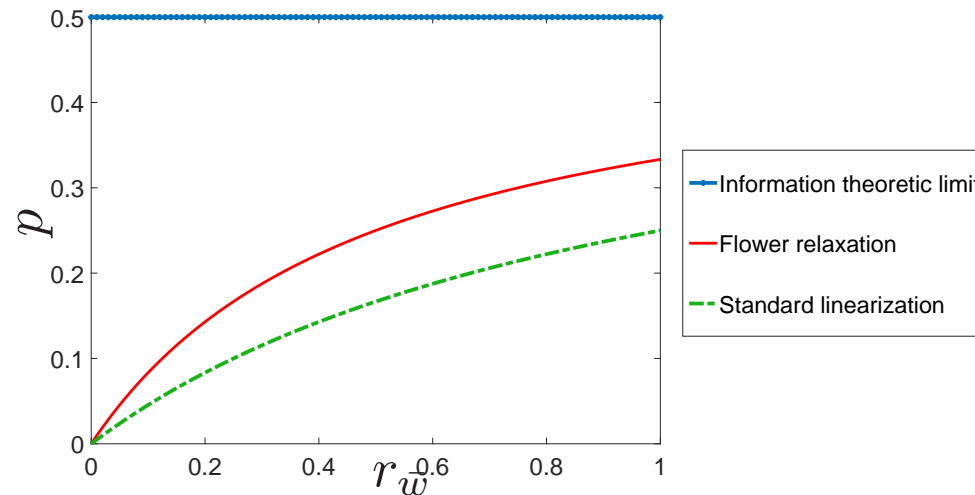
- What is the corruption range, in terms of p , for which any algorithm, **regardless of its computational complexity**, can recover the ground truth with high probability.
- **Theorem (Information theoretic lower bound):** If $p \geq 1/2$, then the probability that rank-one BTF recovers the ground truth is at most $1/2$. Furthermore, if $r_{\bar{x}}, r_{\bar{y}}, r_{\bar{z}}, p$ are positive constants and $p > 1/2$, then with high probability rank-one BTF does not recover the ground truth.
- **Theorem (Information theoretic upper bound):** Assume that $r_{\bar{x}}, r_{\bar{y}}, r_{\bar{z}}$ are positive constants and

$$\lim_{n,m,l \rightarrow \infty} \frac{n + m + l}{\min\{nm, nl, ml\}} = 0.$$

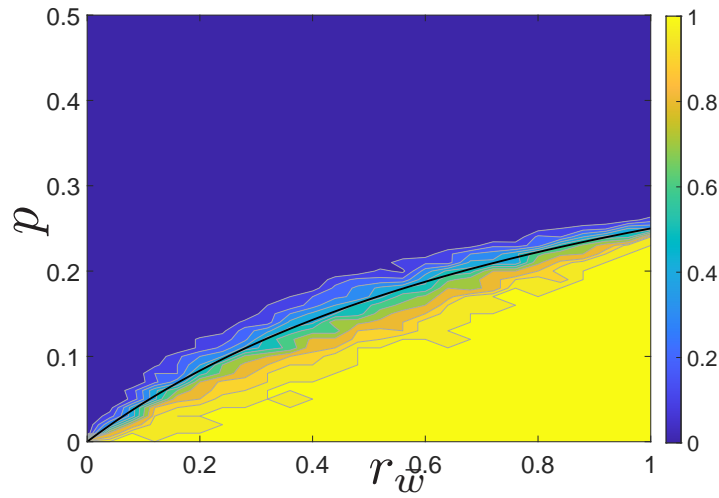
If p is a constant satisfying $p < 1/2$, then rank-one BTF recovers the ground truth with high probability.

Recovery guarantees for LP relaxations

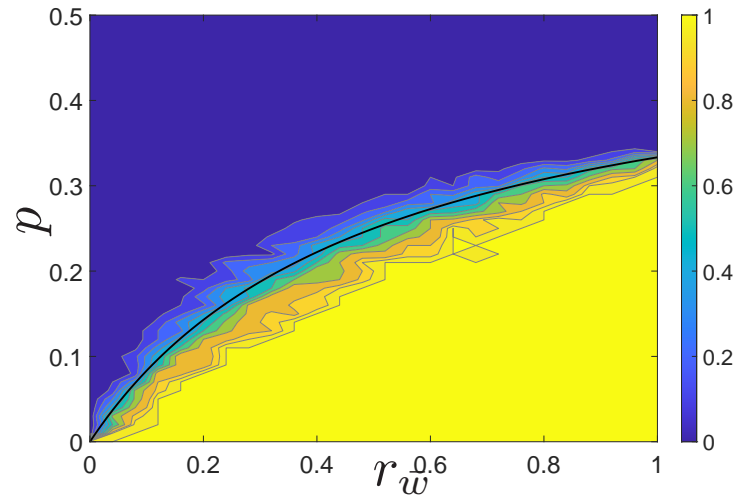
- **Theorem:** Assume that, as $n, m, l \rightarrow \infty$, we have $n \exp(-m)$, $n \exp(-l)$, $m \exp(-n)$, $m \exp(-l)$, $l \exp(-n)$, $l \exp(-m) \rightarrow 0$. If p is a constant satisfying $p < \frac{r\bar{w}}{2(1+r\bar{w})}$, then Problem (sLP) recovers the ground truth with high probability.
- **Theorem:** Assume that, as $n, m, l \rightarrow \infty$, we have $nml \exp(-n)$, $nml \exp(-m)$, $nml \exp(-l) \rightarrow 0$. If p is a constant satisfying $p < \frac{r\bar{w}}{1+2r\bar{w}}$, then Problem (fLP) recovers the ground truth with high probability.



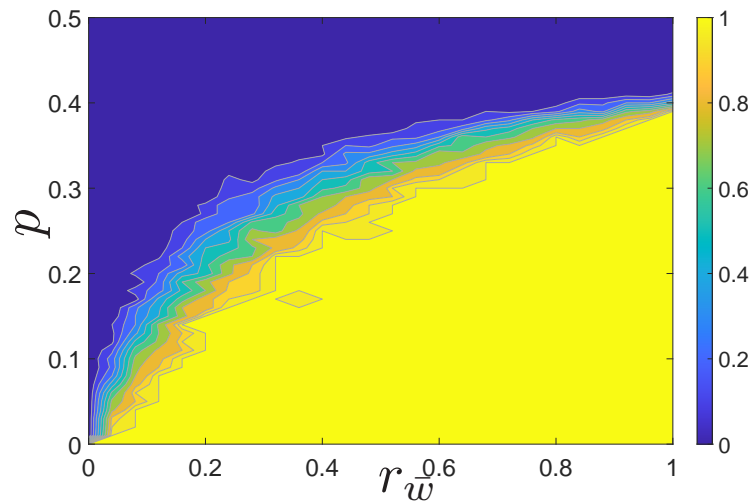
Numerical experiments



(a) Standard LP



(b) Flower LP



(c) Complete LP