

# On the existence of 0/1 polytopes with high semidefinite extension complexity

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## Joint Work with...

### My coauthors:

- Sebastian Pokutta (Georgia Tech)
- Jop Briët (CWI)

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Alternative measure of complexity independent from  $P$  vs  $NP$ .

Lower bounds on the size of *Linear Programs*:

- 1 Any symmetric LP that captures the TSP or Matching polytope must have size  $2^{\Omega(n)}$  [Yannakakis '91]
- 2 There exists a convex hull of 0/1 points that cannot be captured by an LP of size less than  $2^{\Omega(n)}$  [Rothvoss '11]
- 3 Any LP that captures the TSP polytope must have size  $2^{\Omega(n^{1/2})}$  [Fiorini, Massar, Pokutta, Tiwary, de Wolf '12]
- 4 Any LP that  $\rho$ -approximates the Correlation polytope must have size  $2^{\Omega(n/\rho)}$  [Braun, Fiorini, Pokutta, Steurer '12, Braverman, Moitra '13, Pokutta, Braun '13]
- 5 Any LP of relaxation of size  $n^r$  for the Correlation polytope has integrality gap at least as large as  $O(r)$  levels of Sherali-Adams [Chan, Lee, Raghavendra, Steurer '13]

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## What about *Semidefinite Programs*?

Theorem ([Briët, D., Pokutta '13])

*There exists a convex hull of 0/1 points that cannot be captured by an SDP of size less than  $2^{\Omega(n)}$ .*

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Polytope  $P = \{x \in \mathbb{R}^n : Ax \leq b\}$  with  $m$  facets.

### Question

*Can we reduce the number of inequalities needed to define  $P$  by adding auxilliary variables?*

### Definition (Linear Extension)

$Q = \{(z, y) : Cz + Dy = d, y \geq 0, y \in \mathbb{R}^r, z \in \mathbb{R}^l\}$ , is a **linear extension of  $P$  of size  $r$**  if  $\exists \pi : \mathbb{R}^{l+r} \rightarrow \mathbb{R}^n$  such that

$$P = \pi(Q).$$

### Definition (Linear Extension Complexity)

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## Extension Complexity.

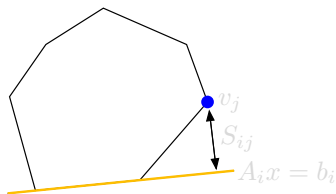
**Some known results (constructions & lower bounds):**

- $\text{xc}(\text{regular } n\text{-gon}) = \Theta(\log n)$  [Ben-Tal, Nemirovski'01]
- $\text{xc}(\text{generic } n\text{-gon}) = \Omega(\sqrt{n})$  [Fironi, Rothvoss, Tiwary'11]
- $\text{xc}(n\text{-permutahedron}) = \Theta(n \log n)$  [Goemans'09]
- $\text{xc}(\text{spanning tree polytope of } K_n) = O(n^3)$  [Kipp-Martin'87]
- $\text{xc}(\text{spanning tree polytope of planar graph } G) = \Theta(n)$   
[Williams'01]
- $\text{xc}(\text{stable set polytope of perfect graph } G) = n^{O(\log n)}$   
[Yannakakis'91]
- ...

## Slack Matrices.

$$\text{Let } A \in \mathbb{R}^{m \times n}, \quad b \in \mathbb{R}^m, \quad \mathcal{V} = \{v_1, \dots, v_N\} \subseteq \mathbb{R}^n \quad \text{s.t.}$$

$$P = \{x \in \mathbb{R}^n \mid Ax \leq b\} = \text{conv}(\mathcal{V})$$



Definition (slack matrix)

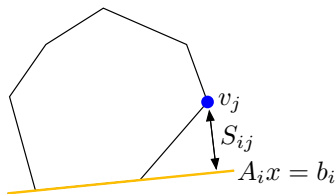
Slack matrix  $S \in \mathbb{R}_+^{m \times N}$  of  $P$  (w.r.t.  $Ax \leq b$  and  $\mathcal{V}$ ):

$$S_{ij} := b_i - A_i v_j, \forall i \in [m], j \in [N]$$

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## Nonnegative Factorizations and Extensions.

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## Definition

A **rank- $r$  nonnegative factorization** of  $S \in \mathbb{R}_+^{m \times n}$  is

$$S = UV \quad \text{where} \quad U \in \mathbb{R}_+^{m \times r} \quad \text{and} \quad V \in \mathbb{R}_+^{r \times n}$$

## Proposition (Extensions from Factorizations)

$Q = \{(x, y) : Ax + Uy = b, y \geq 0\}$   
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## Nonnegative Factorizations and Extensions.

## Definition (Nonnegative Rank)

$$\text{rk}_+(S) := \min\{r \mid \exists \text{ rank-}r \text{ nonnegative factorization of } S\}$$

## Theorem (Factorization Theorem [Yannakakis'91])

For *every* slack matrix  $S$  of  $P$ :

$$\text{xc}(P) = \text{rk}_+(S)$$



## PSD Matrices.

## Definition (PSD matrix)

A matrix  $U \in \mathbb{R}^{r \times r}$  is PSD if  $U$  is symmetric and

$$x^T U x \geq 0 \quad \forall x \in \mathbb{R}^r.$$

Let  $\mathbb{S}_+^r$  denote the set of  $r \times r$  PSD matrices.

## Definition (Spectral Decomposition)

$U$  is  $r \times r$  PSD iff  $U$  admits a *Spectral Decomposition*

$$U = \sum_{i=1}^r \lambda_i u_i u_i^T,$$

$\lambda_1, \dots, \lambda_r \geq 0$ ,  $u_1, \dots, u_r$  an orthonormal basis.

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## Definition (Operator norm)

For a matrix  $T \in \mathbb{R}^{r \times r}$  the operator norm of  $T$  is

$$\|T\|_{\text{op}} = \max_{\|x\|_2=1} \|Tx\|_2$$

For a PSD matrix  $U \in \mathbb{R}^{r \times r}$

$$\|U\|_{\text{op}} = \max_{\|x\|_2=1} x^T U x = \text{largest eigenvalue of } U.$$

## PSD Matrices.

## Definition (Trace)

For a matrix  $T \in \mathbb{R}^{r \times r}$ , we define  $\text{Tr}[T] = \sum_{i=1}^r T_{ii}$ .

## Remark (Trace Inner Product)

For  $A, B \in \mathbb{R}^{r \times r}$  symmetric,  $\text{Tr}[AB] = \sum_{i,j \in [r]} A_{ij}B_{ij}$ .

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For PSD matrices  $U, V \in \mathbb{S}_+^r$ ,

$$\text{Tr}[UV] = \sum_{i,j \in [r]} \lambda_i \gamma_j \langle u_i, v_j \rangle^2 \geq 0,$$

where  $U = \sum_{i=1}^r \lambda_i u_i u_i^\top$  and  $V = \sum_{j=1}^r \gamma_j v_j v_j^\top$  are the respective spectral decompositions.

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Theorem (Factorization Theorem [Gouveia, Thomas, Parrillo '11, Fiorini, Massar, Pokutta, Tiwary, de Wolf '12])

For *every* slack matrix  $S$  of  $P$ :

$$\text{xc}_{\text{psd}}(P) = \text{rk}_{\text{psd}}(S)$$

# Rothvoss's Counting Argument

## Goal:

Show existence of  $X \subseteq \{0, 1\}^n$  with  $\text{xc}(\text{conv}(X)) = 2^{\Omega(n)}$ .

Let  $R = \max_{X \subseteq \{0, 1\}^n} \text{xc}(\text{conv}(X))$ .

## High level:

- 1 “Discretize” an optimal a linear EF for  $\text{conv}(X)$  to *compress the description* of  $X$ .
- 2 Show that the number of discretized linear EFs of size  $R$  is bounded by  $2^{\text{poly}(R, n)}$ .

*Conclusion:*  $2^{2^n}$  subsets of  $\{0, 1\}^n$  means that  $R \geq 2^{\Omega(n)}$ .

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# Discretizing Linear EFs

$$X = \{v_1, \dots, v_N\} \subseteq \{0, 1\}^n. \quad \text{conv}(X) = \{x \in \mathbb{R}^n : Ax \leq b\}.$$

By Hadamard can assume  $A \in \mathbb{Z}^{m \times n}, b \in \mathbb{Z}^m$   
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Theorem ([Briët, D., Pokutta])

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# Rescaling Nonnegative Factorizations

## Lemma

Let  $S \in \mathbb{R}_+^{m \times N}$  with  $\text{rk}_+(S) = r$  and  $\|S\|_\infty = \Delta$ .

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$$\textit{Proof: } U = (u_1 \quad \dots \quad u_r), V = \begin{pmatrix} v_1^\top \\ \vdots \\ v_r^\top \end{pmatrix}$$

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Choose  $\lambda_1, \dots, \lambda_r > 0$  such that  $\|\lambda_i u_i\|_\infty = \|1/\lambda_i v_i\|_\infty$ .

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- Nonnegative setting: can rescale entries of non-negative vector independently and maintain non-negativity.
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**Admissible PSD Rescalings:** For  $A \in \mathbb{R}^{r \times r}$  invertible, send  $U_i \rightarrow A^T U_i A$  and  $V_j \rightarrow A^{-1} V_j A^{-T}$ .

Preserves inner product:

$$\text{Tr}[A^T U_i A A^{-1} V_j A^{-T}] = \text{Tr}[U_i V_j A^{-T} A^T] = \text{Tr}[U_i V_j] = S_{ij}.$$

Map  $U \rightarrow A^T U A$  is a **symmetry of PSD cone**.



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*Proof Idea: Variational Argument*

- 1 Choose rescaling  $A \in \mathbb{S}_+^r$  such that potential  $\max_i \|AU_iA\|_{op} \times \max_j \|A^{-1}V_jA^{-1}\|_{op}$  is minimized.
- 2 Show that if potential  $> r\Delta$ , can find infinitesimal perturbation  $A \rightarrow (I + \epsilon P)A(I + \epsilon P)$  which decreases potential.

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Thank you!