# Nullspace Conditions for Block-Sparse Recovery of Semidefinite Systems

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### **Problem setting**

Consider a linear transformation  $\mathcal{A} \colon \mathcal{S}^n \to \mathbb{R}^m$  given by

 $\mathcal{A}(X) = (A_1 \bullet X, \ldots, A_m \bullet X)^\top,$ 

where  $A_1, \ldots, A_m \in S^n$  are in *block-diagonal form* w.r.t. blocks  $B_1, \ldots, B_k$ .

**Goal:** Finding solutions  $X \succeq 0$  of  $\mathcal{A}(X) = b$  with minimal block support  $BS(X) \coloneqq \{i \in [k] : X_{B_i} \neq 0\}.$ 

## Semidefinite Block-Matrix NSP

**Definition.** A linear transformation  $\mathcal{A}(X)$  in block-diagonal form satisfies the *semidefinite block-matrix null space property* of order *s* iff for all  $S \subset [k], |S| \leq s$  and all  $V \in \text{ker}(\mathcal{A}) \setminus \{0\}$ ,  $V \in S^n$  with  $\lambda_j(V_{B_i}) \leq 0$  for all  $j \in [n_i]$  with  $i \in \overline{S}$ , where  $n_i$  denotes the size of the block  $B_i$ ,

 $\sum_{i\in S}\sum_{j\in [n_i]}\lambda_j(V_{B_i}) < \sum_{i\in \overline{S}}\sum_{j\in [n_i]}|\lambda_j(V_{B_i})|$ 

$$(NSP^*_{*,1,\succeq 0})$$

Formulation as optimization problem:

$$\min\{\|X\|_{*,0}: \mathcal{A}(X) = b, X \succeq 0\}, \qquad (1)$$

with

 $||X||_{*,0} = ||(||X_{B_1}||_{*}, \dots, ||X_{B_k}||_{*})^{\top}||_{0},$ 

where  $\|\cdot\|_*$  denotes the nuclear norm  $\|A\|_* = \sum_{i=1}^n \sigma_i(A)$  with singular values  $\sigma_i(A)$  of A and

 $||x||_0 \coloneqq |\operatorname{supp}(x)| = |\{i \in [n] : x_i \neq 0\}|.$ 

Alternatively, analogous problems without the restriction  $X \succeq 0$  can and have been studied.

Idea: replace non-convex  $\|\cdot\|_{*,0}$  and consider convex relaxation  $\min\{\|X\|_{*,1} : \mathcal{A}(X) = b, X \succeq 0\}$  (2) holds.

Theorem. Let A(X) be a linear transformation in block-diagonal form, b ∈
ℝ<sup>m</sup> and s ≥ 1. The following statements are equivalent:
(i) If A(X) = b has a solution X<sup>0</sup> ∈ S<sup>n</sup><sub>+</sub> with ||X<sup>0</sup>||<sub>\*,0</sub> ≤ s, X<sup>0</sup> is the unique solution of (2).
(ii) A(X) satisfies the semidefinite block-matrix null space property of order

S.

#### **Special Case: Linear Block-Problems**

If  $A_1, \ldots, A_m$  are diagonal matrices, (1) also captures the as of yet unstudied block vector setting for the signed case.

**Corollary.** Choosing (*X*) in diagonal form, above Theorem also shows that block-sparse solutions of linear systems

$$b = Ax = [A[1] \cdots A[k]] x \tag{5}$$
$$x \ge 0,$$

where  $A \in \mathbb{R}^{d \times n}$  consists of k blocks  $A[i] \in \mathbb{R}^{d \times n_i}$  and  $b \in \mathbb{R}^d$ , can successfully be recovered by solving a linear program. More specifically, the following statements for the block-linear system (5) are equivalent: (i) If Ax = b has a nonnegative solution  $x^0 \in \mathbb{R}^n$  with  $||x^0||_{1,0} \leq s$ , then  $x^0$  is the unique solution of

<u>Aim:</u> Characterize, when solving (2) yields a solution of (1) using null space properties (NSPs)
 <u>Motivation:</u> Signal recovery, identifying minimal irreducible subsystems of spectrahedra

#### **Special Case: Linear Problems**

Consider  $1 \times 1$ -blocks. (1) without additional constraint  $X \succeq 0$  reduces to well-studied vector case

$$\min\{\|x\|_0 : Ax = b, x \in \mathbb{R}^n\}.$$
 (3)

Standard result: LP relaxation

$$\min\{\|x\|_1 : Ax = b, x \in \mathbb{R}^n\}$$
 (4)

satisfies the NSP: Any *s*-sparse vector *x* is the unique solution of (4) iff for all  $S \subset [n]$  with  $|S| \leq s$  we have

 $||v_S||_1 < ||v_{\overline{S}}||_1$  for all  $v \in \ker(A) \setminus \{0\}$ .

 $\min\{\|x\|_{1,1}: Ax = b, x \ge 0\}.$ 

(ii) *A* satisfies the nonnegative block-linear null space property of order *s*, i.e.,

$$\sum_{i \in S} \sum_{j \in B_i} v_j < \sum_{i \in \overline{S}} \sum_{j \in B_i} |v_j| \qquad (\text{NSP}_{1,1,\geq 0})$$

holds for all  $S \subset [k]$ ,  $|S| \leq s$  and all  $v \in \text{ker}(A) \setminus \{0\}$ , with  $v_j \leq 0$  for all  $j \in B_i$  with  $i \in \overline{S}$ .

## Outlook

It can be shown that using positive semidefiniteness/nonnegativity captures cases, that can not be recovered by corresponding unrestricted NSPs.

Analogously, (1) and (2) specialize to well studied nonnegative versions of (3) and (4) with corresponding non- negative NSP, see e.g. *Donoho, Foucart/Rauhut*.

It is possible to formulate a more general framework unifying all the above mentioned cases.



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