

# Exploiting Structured Sparsity in Large Scale Semidefinite Programming Problems

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- Kim, Kojima, Mevissen and Yamashita, “Exploiting sparsity in linear and nonlinear inequalities via positive semidefinite matrix completion”, *Mathematical Programming* to appear.

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## Outline

- 1 A simple example for 2 types of sparsities
- 2 Mathematics behind exploiting sparsity

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$$\text{SDP: } \min \sum_{i=1}^{n-1} (X_{ii} + b_i(X_{i,i+1} + X_{i+1,i})) + X_{nn} \quad (1)$$

sub. to (Matrix inequality, diagonal+bordered)

$$M(\mathbf{X}) = \begin{pmatrix} 1 - X_{11} & 0 & \dots & X_{12} \\ 0 & 1 - X_{22} & \dots & X_{23} \\ \dots & \dots & \ddots & \dots \\ X_{21} & X_{32} & \dots & 1 - X_{nn} \end{pmatrix} \succeq \mathbf{O} \quad (2)$$

$$\mathbf{X} = \begin{pmatrix} X_{11} & X_{12} & \dots & X_{1n} \\ X_{21} & X_{22} & \dots & X_{2n} \\ \dots & \dots & \dots & \dots \\ X_{n1} & X_{n2} & \dots & X_{nn} \end{pmatrix} \succeq \mathbf{O} \quad (\text{positive semidefinite})$$

- The number of variables is  $n(n+1)/2$ ;  $X_{ij} = X_{ji}$ .
- domain-space sparsity — Only  $X_{ij}$  ( $|i-j| \leq 1$ ) are used in (1), (2) among all variables  $X_{ij}$  ( $1 \leq i \leq j \leq n$ ).
- range-space sparsity — (2) is diagonal + bordered.

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$$\text{SDP: } \min \sum_{i=1}^{n-1} (X_{ii} + b_i(X_{i,i+1} + X_{i+1,i})) + X_{nn} \quad (1)$$

sub. to (Matrix inequality, diagonal+bordered)

$$M(\mathbf{X}) = \begin{pmatrix} 1 - X_{11} & 0 & \dots & X_{12} \\ 0 & 1 - X_{22} & \dots & X_{23} \\ \dots & \dots & \ddots & \dots \\ X_{21} & X_{32} & \dots & 1 - X_{nn} \end{pmatrix} \succeq \mathbf{O} \quad (2)$$

$$\mathbf{X} = \begin{pmatrix} X_{11} & X_{12} & \dots & X_{1n} \\ X_{21} & X_{22} & \dots & X_{2n} \\ \dots & \dots & \dots & \dots \\ X_{n1} & X_{n2} & \dots & X_{nn} \end{pmatrix} \succeq \mathbf{O} \quad (\text{positive semidefinite})$$

- ↓ conversion with exploiting the domain and range sparsities  
“smaller size” SDP equivalent to the original SDP
- Next, numerical results on the converted SDP

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## Numerical results

- SeDuMi (MATLAB, a primal-dual interior-point method)
- 2.66 GHz Dual-Core Intel Xeon with 12GB memory

size of $X$ = $n$	SeDuMi elapsed time (second)	
	Original SDP	Converted SDP with exploiting d-space & r-space sparsities
10	0.2	0.1
100	1091.4	0.6
1000	-	6.3
10000	-	99.2

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- 1 A simple example for 2 types of sparsities
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## A simple example of positive semidefinite completion problems

Let

$$A(x) = \begin{pmatrix} 3 & 3 & x \\ 3 & 3 & 2 \\ x & 2 & 2 \end{pmatrix} \in \mathbb{S}^3.$$

Question:  $\exists x$  such that  $A(x)$  is positive semidefinite?

Yes, for example,  $A(2) \succeq \mathbf{O}$ . But,  $A(0) \not\succeq \mathbf{O}$ .

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Let  $a_{ij} \in \mathbb{R}$ , and consider

$$A(x_{ij}) = \begin{pmatrix} a_{11} & x_{12} & x_{13} & x_{14} & x_{15} & a_{16} \\ x_{12} & a_{22} & x_{23} & x_{24} & x_{25} & a_{26} \\ x_{13} & x_{23} & a_{33} & a_{34} & x_{35} & a_{36} \\ x_{14} & x_{25} & a_{34} & a_{44} & a_{45} & a_{46} \\ x_{15} & x_{25} & x_{35} & a_{45} & a_{55} & a_{56} \\ a_{16} & a_{26} & a_{36} & a_{46} & a_{56} & a_{66} \end{pmatrix} \in \mathbb{S}^6.$$

Question: How can we check whether  $A(x_{ij})$  is positive semidefinite for some  $x_{ij}$ ?

$\exists x_{ij} \Leftrightarrow$

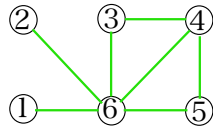
$$A_{\{1,6\}} \equiv \begin{pmatrix} a_{11} & a_{16} \\ a_{61} & a_{66} \end{pmatrix} \succeq \mathbf{O}, A_{\{2,6\}} \succeq \mathbf{O}, A_{\{3,4,6\}} \succeq \mathbf{O}, A_{\{4,5,6\}} \succeq \mathbf{O}.$$

This equivalence holds when the green part has a chordal graph structure.

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$$A(x_{ij}) = \begin{pmatrix} a_{11} & x_{12} & x_{13} & x_{14} & x_{15} & a_{16} \\ x_{12} & a_{22} & x_{23} & x_{24} & x_{25} & a_{26} \\ x_{13} & x_{23} & a_{33} & a_{34} & x_{35} & a_{36} \\ x_{14} & x_{25} & a_{34} & a_{44} & a_{45} & a_{46} \\ x_{15} & x_{25} & x_{35} & a_{45} & a_{55} & a_{56} \\ a_{16} & a_{26} & a_{36} & a_{46} & a_{56} & a_{66} \end{pmatrix} \in \mathbb{S}^6.$$

Sparsity pattern graph  $G(N, E)$



- Each edge  $(i, j)$  is corresponding to an off-diagonal  $a_{ij}$ .
- If the graph  $G(N, E)$  is chordal (i.e. any simple cycle with at least 4 edges has a chord)
- The maximal cliques are  $\{1, 6\}$ ,  $\{2, 6\}$ ,  $\{3, 4, 6\}$ ,  $\{4, 5, 6\}$ .
- Dual of the psd completion problem  $\Rightarrow$  Next

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### Dual of the psd completion problem

- Duality :  $X \succeq O \Leftrightarrow X \bullet Y \equiv \text{trace} X^T Y \geq 0$  for  $\forall Y \succeq O$ .
- A well-known simple case

$$A = \begin{pmatrix} B & O \\ O & C \end{pmatrix} \succeq O \Leftrightarrow B \succeq O, C \succeq O.$$

- Extension

$$A = \begin{pmatrix} B_{11} & B_{12} & O \\ B_{21} & D & C_{12} \\ O & C_{21} & C_{22} \end{pmatrix} \succeq O \Leftrightarrow \begin{cases} \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & D \end{pmatrix} \succeq O, \\ \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix} \succeq O, \\ D + C_{22} = B_{22} \end{cases}$$

- Further extension to a general case  $\Rightarrow$  Next

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$G(N, E)$  : a chordal graph with  $N = \{1, \dots, n\}$  and the max. cliques of  $C_1, \dots, C_\ell$ .  $E^\bullet = E \cup \{(i, i) : i \in N\}$ .

$$\mathbb{S}^n(E^\bullet) = \{Y \in \mathbb{S}^n : Y_{ij} = 0 \text{ if } (i, j) \notin E^\bullet\}.$$

$$\mathbb{S}_+^C = \{Y \succeq O : Y_{ij} = 0 \text{ if } (i, j) \notin C \times C \text{ for } \forall C \subseteq N\}.$$

**Theorem** (Agler, Helton, McCulough and Rodman 1988)

Suppose  $M \in \mathbb{S}^n(E^\bullet)$ .  $M \succeq O$  iff

$$M = Y^1 + Y^2 + \dots + Y^\ell \text{ for } \exists Y^k \in \mathbb{S}_+^{C_k} \text{ (} k = 1, \dots, \ell \text{)}.$$

$$\textcircled{1} - \textcircled{2} - \textcircled{3} \quad C_1 = \{1, 2\}, C_2 = \{2, 3\}. \quad M : \mathbb{R}^m \rightarrow \mathbb{S}^3(E^\bullet).$$

$$M(u) = \begin{pmatrix} M_{11}(u) & M_{12}(u) & 0 \\ M_{21}(u) & M_{22}(u) & M_{23}(u) \\ 0 & M_{32}(u) & M_{33}(u) \end{pmatrix}$$

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$$M(u) \succeq O \quad M(u) = \begin{pmatrix} Y_{11}^1 & Y_{12}^1 & 0 \\ Y_{12}^1 & Y_{22}^1 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 0 & Y_{22}^2 & Y_{23}^2 \\ 0 & Y_{32}^2 & Y_{33}^2 \end{pmatrix}$$

$$\Leftrightarrow \begin{cases} M_{11} = Y_{11}^1, M_{12} = Y_{12}^1, \\ M_{22} = Y_{22}^1 + Y_{22}^2, \\ M_{23} = Y_{23}^2, M_{33} = Y_{33}^2, \\ \square \succeq O, \square \succeq O \end{cases} \Leftrightarrow \begin{cases} \begin{pmatrix} M_{11}(u) & M_{12}(u) \\ M_{21}(u) & Y_{22}^1 \end{pmatrix} \succeq O, \\ \begin{pmatrix} M_{22}(u) - Y_{22}^1 & M_{23}(u) \\ M_{32}(u) & M_{33}(u) \end{pmatrix} \succeq O \end{cases}$$

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