

On positive duality gaps in semidefinite programming

Gábor Pataki

UNC Chapel Hill

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A pair of Semidefinite Programs (SDP)

$$\begin{array}{ll} \sup_x c^T x & \inf_Y B \bullet Y \\ (P) \quad s.t. \sum_{i=1}^m x_i A_i \preceq B & Y \succeq 0 \quad (D) \\ & A_i \bullet Y = c_i \forall i. \end{array}$$

Here

- A_i, B are symmetric matrices, $c, x \in \mathbb{R}^m$.
- $A \preceq B$ means that $B - A$ is symmetric positive semidefinite (psd).
- $A \bullet B = \sum_{i,j} a_{ij} b_{ij}$.

SDP duality

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Ideally: $\exists x^*, \exists Y^* : c^T x^* = B \bullet Y^*$.

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But: **pathologies** occur, as nonattainment, positive gaps.

→ in such cases we cannot certify optimality.

Example: positive duality gap

Primal:

$$\begin{array}{l} \sup x_2 \\ \text{s.t. } x_1 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \preceq \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \end{array}$$

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$x_2 = 0$ identically \Rightarrow primal = 0.

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Dual matrix $Y \succeq 0$

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Dual matrix $Y \succeq 0$

$$\text{1st dual constraint } \Rightarrow y_{11} = 0 \Rightarrow Y = \begin{pmatrix} 0 & 0 & 0 \\ 0 & y_{22} & y_{23} \\ 0 & y_{23} & y_{33} \end{pmatrix}$$

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$$\text{2nd dual constraint } \Rightarrow y_{22} = 1 \Rightarrow \text{dual opt} = 1$$

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Looks quite odd: x_1 “only exists” to create a zero block in the dual matrix.

Positive gaps

- Maybe the “worst/most interesting” pathology.
- Solvers fail, or report a wrong solution.
- Good model of positive gaps in more general convex programs

Literature

- Pathological semidefinite **systems P, 2011** –
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- Pathological semidefinite **systems P, 2011** –
“Bad semidefinite programs: they all look the same”
- It does not distinguish among bad objective functions.
- Positive gaps related to complementarity in homogeneous systems: **Tuncel, Wolkowicz, 2012**
- Weak infeasibility: **Lourenco, Muramatsu, Tsuchiya, 2014**
- Infeasibility, weak infeasibility: **Liu, P 2015, 2017**

Literature: how to solve some pathological SDPs

- Facial reduction of Borwein-Wolkowicz, Waki-Muramatsu, Pataki: implemented by **Permenter, Parrilo 2014;**
Permenter, Friberg, Andersen 2015
- Very simple facial reduction (just inspect the constraints):
Zhu, P, Tran Dinh (Sieve-SDP) 2017
- SPECTRA, exact arithmetic SDP solver
Henrion-Naldi-El Din 2016
- Douglas-Rachford splitting:
Liu, Ryu, Yin 2017
- Homotopy method
Hauenstein, Liddell, Zhang 2018

Main ideas

- Look at small instances.
- **Proposition:** positive gap $\Rightarrow m \geq 2$.
- Fully characterize the $m = 2$ case.

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- Fully characterize the $m = 2$ case.
- Show that the $m = 2$ case sheds light on larger m .
- Precisely: the structure that causes positive gap when $m = 2$ often does the same when $m > 2$.

Main ideas

- Look at small instances.

Proposition: positive gap $\Rightarrow m \geq 2$.

- Fully characterize the $m = 2$ case.
- Show that the $m = 2$ case sheds light on larger m .
- Precisely: the structure that causes positive gap when $m = 2$ does the same in many cases even if $m > 2$.
- Reformulate
- Borrow ideas from linear system of equations:
to show $Ax = b$ is infeasible, we create an equation
 $\langle 0, x \rangle = 1$.

Recall: a pair of Semidefinite Programs (SDPs)

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Reformulations of $(P) - (D)$ are obtained by

- Choose T invertible, and

$$B \leftarrow T^T B T, A_i \leftarrow T^T A_i T \forall i.$$

- Elementary row operations on (D) : e.g., exchange two constraints $A_i \bullet Y = c_i$ and $A_j \bullet Y = c_j$.
- Choose $\mu \in \mathbb{R}^m$ and

$$B \leftarrow B + \sum_{i=1}^m \mu_i A_i.$$

Reformulations preserve positive gaps (if any).

Suppose $m = 2$.

Then positive gap $\Leftrightarrow \exists$ reformulation

sup $c'_2 x_2$

$$s.t. \ x_1 \begin{pmatrix} \Lambda & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \end{pmatrix} + x_2 \begin{pmatrix} \times & \times & \times & M \\ \hline \times & \Sigma & & \\ \hline \times & & -I_s & \\ \hline M^T & & & \end{pmatrix} \preceq \begin{pmatrix} I_p & & & \\ \hline & I_{r-p} & & \\ \hline & & & \\ \hline & & & \end{pmatrix}$$

where $\Lambda \succ 0$, $M \neq 0$, $c'_2 > 0$, $s \geq 0$.

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Proof of \Leftarrow

This is the easy direction.

Essentially reuse the argument from before.

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where $\Lambda \succ 0$, $M \neq 0$, $c'_2 > 0$, $s \geq 0$.

Proof of $\Leftarrow M \neq 0 \Rightarrow x_2 = 0 \Rightarrow \text{primal} = 0$.

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Proof of \Leftarrow Dual matrix $Y \succeq 0$

1st dual constraint $\Rightarrow \Lambda \bullet Y(1:p, 1:p) = 0$

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\Rightarrow 1st p rows and columns of Y are zero.

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where $\Lambda \succ 0$, $M \neq 0$, $c'_2 > 0$, $s \geq 0$.

$$\Rightarrow \text{dual is}$$

$$s.t. \ \begin{pmatrix} I_{r-p} & 0 \\ 0 & 0 \end{pmatrix} \bullet Y'$$

$$\begin{pmatrix} \Sigma & 0 \\ 0 & -I_s \end{pmatrix} \bullet Y' = c'_2 > 0$$

$$Y' \succeq 0,$$

Suppose $m = 2$.
 Then positive gap $\Leftrightarrow \exists$ reformulation

$$\begin{array}{l}
 \sup c'_2 x_2 \\
 \text{s.t. } x_1
 \end{array}
 \left(\begin{array}{c|c|c|c}
 \Lambda & & & \\
 \hline
 & & & \\
 \hline
 & & & \\
 \hline
 & & &
 \end{array} \right)
 + x_2
 \left(\begin{array}{c|c|c|c}
 \times & \times & \times & M \\
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 \times & \Sigma & & \\
 \hline
 \times & & -I_s & \\
 \hline
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 \end{array} \right)
 \stackrel{\sim}{=}
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 \hline
 & & & \\
 \hline
 & & &
 \end{array} \right)$$

where $\Lambda \succ 0$, $M \neq 0$, $c'_2 > 0$, $s \geq 0$.

\Rightarrow dual optimal value > 0 .

Simple **certificate** of the positive gap

When does the underlying system admit a gap?

Given

$$(P_{SD}) \quad \sum_{i=1}^m x_i A_i \preceq B$$

is there $c \in \mathbb{R}^m$ such that there is a positive gap?

Suppose $m = 2$. Then $\exists(c_1, c_2)$ with positive gap
 $\Leftrightarrow (P_{SD})$ has reformulation

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where $\Lambda \succ 0$, $M \neq 0$, $s \geq 0$.

How about $m > 2$?

Similar example with $m = 3$

sup x_3

$$s.t. \ x_1 \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 & 1 \\ 1 & 0 \\ 0 & 0 \end{pmatrix} + x_3 \begin{pmatrix} 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \preceq \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \end{pmatrix}.$$

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Dual : Variable $Y = (y_{ij}) \succeq 0$

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1st two dual constraints \Rightarrow 1st two rows and columns of Y are zero.

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Primal = 0.

\Rightarrow Dual is equivalent to:

$$\begin{aligned} \inf \quad & \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \bullet Y' \\ s.t. \quad & \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \bullet Y' = 1 \\ & Y' \succeq 0, \end{aligned}$$

Same structure as in the 2 variable case

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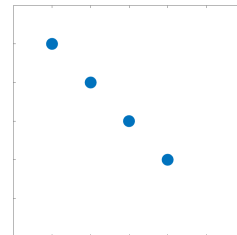
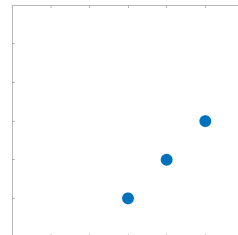
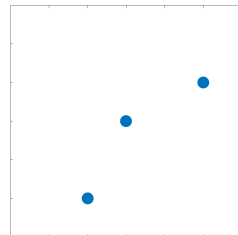
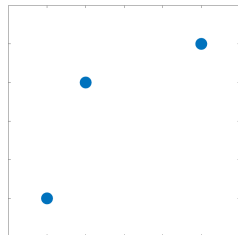
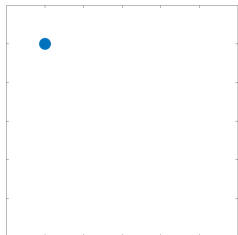
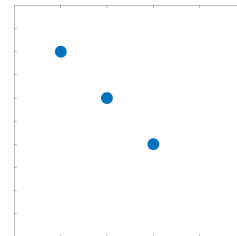
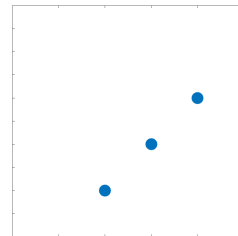
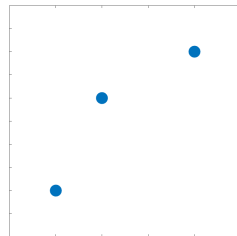
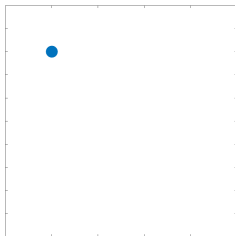
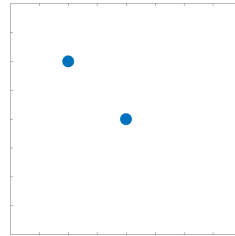
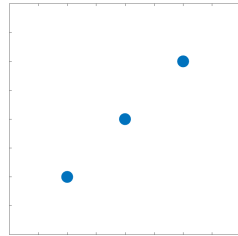
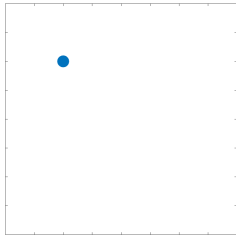
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Same structure as in the 2 variable case

We can create such an instance for any
 $m = 2, 3, 4, \dots$ with $n = m + 1$

Sparsity structure when $m = 2, 3, 4$ ($n = m + 1$)



How do we get these instances? Background: facial reduction

Given H affine subspace, K closed convex cone, a facial reduction algorithm (FRA) works as:

- (1) If $\text{ri } K \cap H = \emptyset$, find $y \in H^\perp \cap (K^* \setminus K^\perp)$.
- (2) Replace K by $K \cap y^\perp$. Goto (1).

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Facial reduction sequence:

The sequence y_1, y_2, \dots generated by the FRA.

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Facial reduction sequence:

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Singularity degree:

Is the smallest number of FRA steps, until the FRA stops.

Back to $m = 2$ example:

$$\begin{array}{l} \text{sup } x_2 \\ \text{s.t. } x_1 \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{A_1} + x_2 \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{A_2} \preceq \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_B \end{array}$$

Here (A_1) is a facial reduction sequence for (D) .

Back to $m = 2$ example:

$$\begin{array}{l} \text{sup } x_2 \\ \text{s.t. } x_1 \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{A_1} + x_2 \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{A_2} \preceq \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_B \end{array}$$

Here (A_1) is a facial reduction sequence for (D) .

$A_1 \bullet Y = 0$ proves that dual matrix must look like

$$Y = \begin{pmatrix} 0 & 0 & 0 \\ 0 & y_{22} & y_{23} \\ 0 & y_{23} & y_{33} \end{pmatrix}$$

Back to larger example:

sup x_3

$$\begin{array}{l}
 s.t. \ x_1 \underbrace{\begin{pmatrix} 1 & & & \\ & 0 & & \\ & & 0 & \\ & & & 0 \end{pmatrix}}_{A_1} + x_2 \underbrace{\begin{pmatrix} 0 & & 1 & \\ & 1 & & \\ & & 0 & \\ 1 & & & 0 \end{pmatrix}}_{A_2} + x_3 \underbrace{\begin{pmatrix} 0 & & & \\ & 0 & 1 & \\ & & 1 & \\ 1 & & & 0 \end{pmatrix}}_{A_3} = \underbrace{\begin{pmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 0 \end{pmatrix}}_B
 \end{array}$$

Here (A_1, A_2) is a facial reduction sequence for (D) .

Back to larger example:

sup x_3

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Here (A_1, A_2) is a facial reduction sequence for (D) .

$A_1 \bullet Y = A_2 \bullet Y = 0$ proves that dual matrix must look like

$$Y = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & y_{33} & y_{34} \\ 0 & 0 & y_{43} & y_{44} \end{pmatrix}$$

Theorem:

- $\text{sing}(D) \leq m.$

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- $\text{sing}(\mathbf{D}) \leq m$.
- $= m \Rightarrow$ no gap. (Easy)
- $\forall m \geq 2, \forall g > 0 \exists$ instance s.t.
 $\text{sing}(\mathbf{D}) = m - 1$ and gap is g . (See examples)

Analogous results for the homogeneous dual

$$A_i \bullet Y = 0 \quad \forall i$$

$$B \bullet Y = 0 \quad (HD)$$

$$Y \succeq 0$$

See paper ...

Problem library and computational results

- Instances with $m = 2, 3, \dots, 11$; gap = 10.

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 - *gap_single_finite_messy_m*
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- **Messy** means we applied a similarity transformation $T^T()T$.

Results

	GAP, SINGLE, FINITE		GAP, SINGLE, INFINITE	
	CLEAN	MESSY	CLEAN	MESSY
MOSEK	1	1	0	0
SDPA-GMP	1	1	0	0
PP+MOSEK	10	1	10	0
SIEVE-SDP + MOSEK	10	1	10	0

- PP: preprocessor of Permenter and Parrilo
- Sieve-SDP: preprocessor of Zhu, Pataki, Tran-Dinh

Note: SPECTRA works when $m = 2$.

Summary:

- Positive gaps: “worst/most interesting” pathology of SDPs.
- Complete characterization for $m = 2$ by reformulation.
- Complete characterization of positive gap systems with $m = 2$.
- Similarly structured positive gap SDPs in any dimension.
- Highest singularity degree that leads to a positive gap.
- Challenging problem library.

Thank you!