

Combinatorial certificates in SDP duality:  
how elementary row operations help

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# SDP duality

$$\begin{array}{ll} \inf_X C \bullet X & \sup_y b^T y \\ (P) \quad s.t. \quad X \succeq 0 & \sum_{i=1}^m y_i A_i \preceq C \quad (D) \\ & A_i \bullet X = b_i \quad (i \in [m]). \end{array}$$

**Easy:** If  $X$  and  $y$  are feasible, then  $C \bullet X \geq b^T y$ .

**But:**

- unattained optimal values
- positive duality gaps
- Slater's condition helps, but it can fail.
- poly time solvable only under restrictive assumptions

## Alternative system (Farkas' lemma) to prove infeasibility

$$\begin{array}{ll} \inf_X C \bullet X & \sup_y b^T y \\ (P) \quad s.t. \quad X \succeq 0 & \sum_{i=1}^m y_i A_i \succeq 0 \quad (ALT-P) \\ & A_i \bullet X = b_i \quad (i \in [m]). \quad b^T y = -1 \end{array}$$

Easy:  $(P)$  infeasible  $\Leftrightarrow (ALT-P)$  feasible

But:

- $\nRightarrow$
- Thus  $(ALT-P)$  is not useful for complexity
- not known whether feasibility of  $(P)$  is in P

## Some partial history of duality in conic linear programs:

- **Duffin, 1956:** Infinite programs
- **Berman, 1973:** Cones, matrices, and mathematical programming
- **Ben-Israel, Charnes, Kortanek, 1968:** Duality and asymptotic solvability over cones
- **Berman, Ben-Israel, 1971:** Duality and asymptotic solvability over cones
- **:**

## Better understanding in the last 15+ years based on:

- (1) On the closedness of the linear image of a closed convex cone, P, 2007
- (2) Facial reduction: Borwein-Wolkowicz 80-81; Waki-Muramatsu 2012; P 2000, 2013, Liu-P 2016
- (3) Elementary row operations

## Elementary row operations for linear systems:

$Ax = b$  is infeasible  $\Leftrightarrow$  by elementary row operations it can be transformed into

$$A'x = b'$$

$$0^T x = 1$$

- $\Leftarrow$  is easy, so a good certificate
- Can be used to generate any infeasible system

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We will use these operations to understand SDPs:

**Part 1** Badly behaved feasible SDPs

**Part 2** Infeasible SDPs

**Part 3** Asymptotic feasibility, weak infeasibility

**Part 4** Ramana's perfect infeasibility certificate

# Reformulations

$$\begin{array}{ll} \inf_X C \bullet X & \sup_y b^T y \\ (P) \quad s.t. \quad X \succeq 0 & \sum_{i=1}^m y_i A_i \preceq C \quad (D) \\ & A_i \bullet X = b_i \quad (i \in [m]). \end{array}$$

We obtain a reformulation of (P) – (D) by a sequence of the following:

- Elementary row operations on the equations of (P)
- For  $\lambda \in \mathbb{R}^m$

$$C \leftarrow C + \sum_i \lambda_i A_i$$

- For a  $V$  invertible matrix

$$A_i \leftarrow V^T A_i V \quad (i \in [m]), \quad C \leftarrow V^T C V,$$

## Part 1: Why all bad SDPs look the same

Example

$$\begin{array}{l} \sup 2y_1 \\ \text{s.t. } y_1 \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \preceq \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \end{array}$$

Only feasible  $y_1$  is  $y_1 = 0$ .

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Only feasible  $y_1$  is  $y_1 = 0$ .

Primal: Variable is  $X \succeq 0$ . (P) equivalent to

$$\begin{aligned} & \inf x_{11} \\ & s.t. \quad x_{11}, x_{22} \geq 0, x_{11}x_{22} \geq 1 \end{aligned}$$

Hyperbola asymptote: unattained 0 optimal value.

## Part 1: Why all bad SDPs look the same

- Semidefinite system:

$$(D_{sys}) \quad \sum_{i=1}^m y_i A_i \preceq C$$

We say  $(D_{sys})$  is badly behaved, if  $\exists b$  s.t.

$$\sup\{ b^\top y \mid y \in D_{sys} \}$$

is finite, but  $(P)$  has no solution with the same value.

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is finite, but  $(P)$  has no solution with the same value.

- A **slack matrix** in  $(D)$  is a matrix

$$Z := B - \sum_{i=1}^m y_i A_i \succeq 0.$$

## Part 1: Why all bad SDPs look the same

One variable case: In

$$(D_{sys}) \quad y_1 \begin{pmatrix} \overbrace{V_{11}}^r & V_{12} \\ V_{12}^T & \underbrace{V_{22}}_{n-r} \end{pmatrix} \preceq \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix},$$

suppose rhs is the maximum rank slack. Then  $(D_{sys})$  is badly behaved  $\Leftrightarrow$

$$V_{22} \succeq 0, \quad R(V_{12}^T) \not\subseteq R(V_{22}).$$

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So, just a larger version of

$$y_1 \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \preceq \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

## Part 1: Why all bad SDPs look the same

**Theorem:** Assume in

$$(D_{sys}) \quad \sum_{i=1}^m y_i A_i \preceq C$$

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Then  $(D_{sys})$  is badly behaved  $\Leftrightarrow \exists$  reformulation

$$(D'_{sys}) \quad \sum_{i=1}^m y_i A'_i \preceq C$$

s.t.

$$y_1 A'_1 \preceq C$$

is badly behaved.

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s.t.

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is badly behaved.

I.e., one variable bad subsystem certifies bad behavior.

## Part 1: Why all bad SDPs look the same

Example

$$y_1 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + y_2 \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \preceq \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

has positive gap, when we seek  $\sup y_2$ .

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One variable bad subsystem:

$$\cancel{y_1} \begin{pmatrix} \cancel{1} & \cancel{0} & \cancel{0} \\ \cancel{0} & \cancel{0} & \cancel{0} \\ \cancel{0} & \cancel{0} & \cancel{0} \end{pmatrix} + y_2 \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \preceq \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

## Part 2: Infeasibility

### Example

$$\underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{A_1} \bullet X = \underbrace{0}_{b_1}$$
$$\underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{A_2} \bullet X = \underbrace{-1}_{b_2}$$
$$X \succeq 0$$

Farkas lemma certificate fails: there is no  $y \in \mathbb{R}^2$  s.t.

$$y_1 A_1 + y_2 A_2 \succeq 0, \quad y_1 b_1 + y_2 b_2 = -1.$$

## Part 2: Infeasibility

### Example

$$\begin{array}{l} \left( \begin{array}{ccc} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{array} \right) \bullet X = 0 \\ \left( \begin{array}{ccc} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{array} \right) \bullet X = -1 \\ X \succeq 0 \end{array}$$

Still, it is easy to see why it is infeasible:

- Suppose  $X$  feasible  $\Rightarrow x_{11} = 0$   
 $\Rightarrow x_{12} = x_{13} = 0$   
 $\Rightarrow x_{22} = -1$ , contradiction!

We will find such a structure in every infeasible semidefinite system.

Theorem: (P) infeasible  $\Leftrightarrow$  it has a reformulation

$$\begin{aligned}
 A'_i \bullet X &= 0 \quad (i = 1, \dots, k) \\
 A'_{k+1} \bullet X &= -1 \\
 &\vdots \\
 X &\succeq 0
 \end{aligned}
 \tag{P}_{\text{ref}}$$

where  $k \geq 0$ , and for  $i = 1, \dots, k + 1$

$$A'_1 = \begin{pmatrix} \overbrace{I}^{r_1} & \overbrace{0}^{n-r_1} \\ \mathbf{0} & \mathbf{0} \end{pmatrix}, \quad A'_i = \begin{pmatrix} \overbrace{\times}^{r_1+\dots+r_{i-1}} & \overbrace{\times}^{r_i} & \overbrace{\times}^{n-r_1-\dots-r_i} \\ \times & I & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix}$$

with  $r_i \geq 0 \forall i$ .

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with  $r_i \geq 0 \forall i$ . Liu-P, 2015

Proof of “ $\Leftarrow$ ” : Suppose  $X$  feasible in  $(\text{P}_{\text{ref}})$

$\Rightarrow$  first  $r_1$  rows of  $X$  are 0

...

$\Rightarrow$  first  $r_1 + \dots + r_k$  rows of  $X$  are 0

$\Rightarrow$  trace of diag. block of  $X$  is  $-1$ , contradiction!

Proof of “ $\Rightarrow$ ” : Also not hard P-Touzov, 2022

Theorem: (P) infeasible  $\Leftrightarrow$  it has a reformulation

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with  $r_i \geq 0 \forall i$ . Liu-P, 2015

A normal form of empty spectrahedra

## Application 1: generating infeasible SDPs

By this theorem, we can generate **any** infeasible SDP, as:

- (1) Generate a system like  $(P_{\text{ref}})$
- (2) Reformulate it.

There are only finitely many representative infeasible SDPs!

How many? # of partitions of  $\{1, \dots, n\}$  into nonempty subsets

Problem library by **Liu-P** 2016: infeasible SDPs

## Application 2: better understanding of polynomial optimization

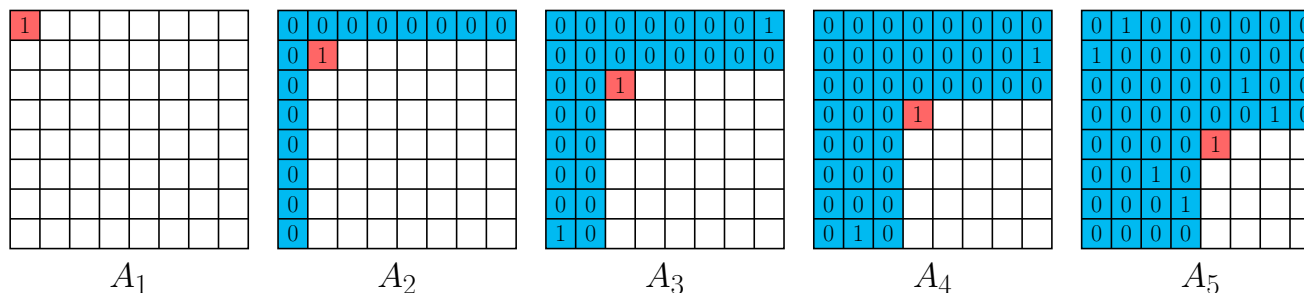
**Example: Motzkin polynomial**

$$f(x, y) = 1 - 3x^2y^2 + x^2y^4 + x^4y^2$$

It is  $\geq 0$ , but not a sum of squares (SOS)

Checking SOSness is an SDP (Lasserre, Parrilo)

The SDP is infeasible, and in the form of  $(P_{\text{ref}})$  without any reformulation

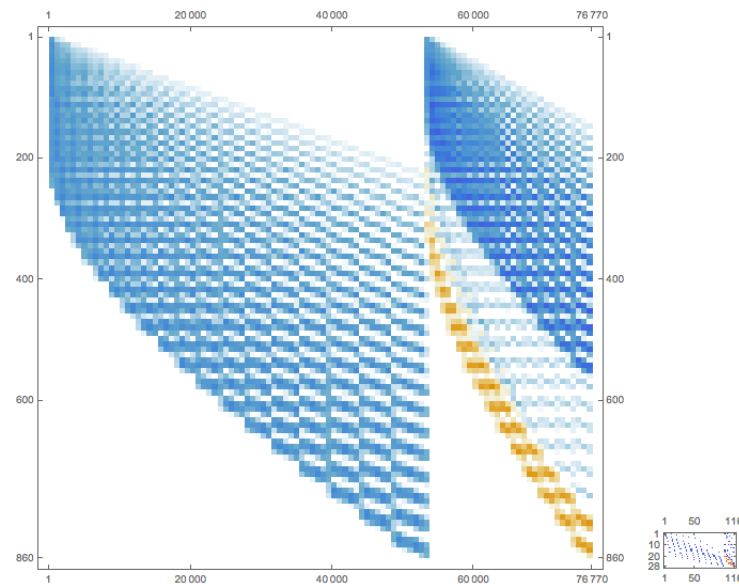


with  $(b_1, b_2, b_3, b_4, b_5) = (0, 0, 0, 0, -3)$

Paper: P-Touzov, 2022, FOCM

## Application 3: recognizing infeasibility in practice

- Sometimes we do not even have to reformulate an SDP to find the trivial structure that proves infeasibility ... or to reduce the SDP.
- **Zhu–P–Tran-Dinh, 2018** Fast Sieve-SDP preprocessor
- Before and after picture of an SDP



Though this is a best case example, many polynomial optimization problems get greatly reduced.

## Part 3: Asymptotic feasibility, weak infeasibility

- Recall dual, without objective function:

$$(D) \quad \sum_{i=1}^m y_i A_i \preceq C$$

- We say  $(D)$  is asymptotically feasible, if an arbitrarily small perturbation of  $C$  makes it feasible.
- Feasible  $\Rightarrow$  asymptotically feasible. But:  $\nRightarrow$

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$$(D) \quad \sum_{i=1}^m y_i A_i \preceq C$$

- We say  $(D)$  is asymptotically feasible, if an arbitrarily small perturbation of  $C$  makes it feasible.
- Feasible  $\Rightarrow$  asymptotically feasible. But:  $\nRightarrow$
- We say  $(D)$  is weakly infeasible, if infeasible, but asymptotically feasible.
- Weakly infeasible SDPs are hard: often mistaken for feasible ones by solvers

## Part 3: Asymptotic feasibility, weak infeasibility

Example:

$$y_1 \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{A_1} + y_2 \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{A_2} \not\preceq \underbrace{\begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}}_C$$

is weakly infeasible.

Infeasible:

$$S := C - y_1 A_1 - y_2 A_2 = \begin{pmatrix} -y_1 & 0 & -y_2 \\ 0 & -y_2 & 1 \\ -y_2 & 1 & 0 \end{pmatrix} \not\preceq 0$$

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is weakly infeasible.

Asymptotically feasible: for any  $\epsilon > 0$

$$S_\epsilon := C - y_1 A_1 - y_2 A_2 = \begin{pmatrix} -y_1 & 0 & -y_2 \\ 0 & -y_2 & 1 \\ -y_2 & 1 & \epsilon \end{pmatrix} \succ 0 :$$

choose  $y_2$  to make lower right  $2 \times 2$  block  $\succ 0$ ; then choose  $y_1$  to make  $S_\epsilon \succ 0$ .

Theorem: (D) asymptotically feasible  $\Leftrightarrow$  it has a reformulation

$$\sum_{i=1}^m y_i A'_i \preceq C'$$

where  $0 \leq \ell \leq m$  and for  $i = 1, \dots, \ell$

$$A'_1 = \begin{pmatrix} \overbrace{I}^{r_1} & \overbrace{0}^{n-r_1} \\ \mathbf{0} & \mathbf{0} \end{pmatrix}, \quad A'_i = \begin{pmatrix} \overbrace{\times}^{r_1+\dots+r_{i-1}} & \overbrace{\times}^{r_i} & \overbrace{\times}^{n-r_1-\dots-r_i} \\ \times & I & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix},$$

and

$$C' = \begin{pmatrix} \overbrace{\times}^{r_1+\dots+r_\ell} & \overbrace{\times}^{r_{\ell+1}} & \overbrace{\times}^{n-r_1-\dots-r_{\ell+1}} \\ \times & I & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix}$$

with  $r_i \geq 0 \forall i$ .

Lourenco-Muramatsu-Tsuchiya, 2014 essentially same statement, Liu-P, 2017 stated in terms of reformulations, and short proof

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with  $r_i \geq 0 \forall i$ .

Proof of " $\Leftarrow$ ": Just like in special case

Proof of " $\Rightarrow$ ": (D) asymptotically feasible  $\Leftrightarrow$  its Farkas' lemma system

$$A_i \bullet X = 0 \forall i, \quad C \bullet X = -1, \quad X \succeq 0$$

is infeasible. + Reformulate this system.

## Part 3: Asymptotic feasibility, weak infeasibility

- We can define asymptotic feasibility, weak infeasibility of  $(P)$  similarly.
- Combining certificates for infeasibility and asymptotic feasibility  $\rightarrow$  generating algorithm for **all** weakly infeasible SDPs.  
**P, Touzov 2022**

## Part 4: Ramana's alternative system

**Ramana, 1995:** A perfect dual for  $(P)$  or  $(D)$  : no duality gap, always attains optimal value

Gives an exact alternative system for  $(D)$  and best complexity results to date.

I will talk about the alternative system for  $(P)$ .

## Part 4: Ramana's alternative system

Motivation: LP infeasibility

$$\begin{array}{ll} (P) & Ax = b \\ & x \geq 0 \end{array} \qquad \begin{array}{ll} & A^T y \geq 0 \\ & b^T y = -1 \end{array} \quad (ALT-P)$$

Three desirable properties:

- (1) Data of  $(P)$  is same as that of  $(ALT-P)$  ? **YES**
- (2)  $(P)$  is infeasible  $\Leftrightarrow (ALT-P)$  is feasible? **YES**
- (3) When feasible, they have poly size solutions ? **YES**

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$$\begin{array}{ll} (LP-P) & Ax = b \\ & x \geq 0 \end{array} \qquad \begin{array}{ll} & A^T y \geq 0 \\ & b^T y = -1 \end{array} \quad (ALT-P)$$

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- (2)  $(P)$  is infeasible  $\Leftrightarrow (ALT-P)$  is feasible? **YES**
- (3) When feasible, they have poly size solutions ? **YES**

**Corollary:** LP feasibility  $\in \mathbf{NP} \cap \mathbf{co-NP}$  in both Turing and real number model. Known already in the sixties.

## Part 4: Ramana's alternative system

Now to SDP

$$\mathcal{A}X := (A_1 \bullet X, \dots, A_m \bullet X)^\top, \quad \mathcal{A}^*y := \sum_{i=1}^m y_i A_i$$

$$\begin{array}{ll} (P) & \mathcal{A}X = b \\ & X \succeq 0 \end{array} \qquad \begin{array}{ll} & \mathcal{A}^*y \succeq 0 \\ & b^T y = -1 \end{array} \quad (ALT-P)$$

$(ALT-P)$  is traditional alternative system

Three desirable properties:

- (1) Data of  $(P)$  is same as that of  $(ALT-P)$  ? **YES**
- (2)  $(P)$  is infeasible  $\Leftrightarrow (ALT-P)$  is feasible? **NO:  $\Rightarrow$  may fail.**
- (3) When feasible, they have poly size solutions ? **NO**

## Part 4: Ramana's alternative system

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Three desirable properties:

- (1) Data of  $(P)$  is same as that of  $(ALT-P)$  ? **YES**
- (2)  $(P)$  is infeasible  $\Leftrightarrow (ALT-P)$  is feasible? **NO**
- (3) When feasible, they have poly size solutions ? **NO**

**Corollary:**  $(ALT-P)$  is not useful for complexity

## Part 4: Ramana's alternative system (Ram-Alt-P)

Three desirable properties:

- (1) Data of  $(P)$  is same as that of (Ram-Alt-P) ? **YES**
- (2)  $(P)$  is infeasible  $\Leftrightarrow$  (Ram-Alt-P) is feasible? **YES**
- (3) When feasible, they have poly size solutions ? **NO**

## Part 4: Ramana's alternative system (Ram-Alt-P)

Three desirable properties:

- (1) Data of  $(P)$  is same as that of (Ram-Alt-P) ? YES
- (2)  $(P)$  is infeasible  $\Leftrightarrow$  (Ram-Alt-P) is feasible? YES
- (3) When feasible, they have poly size solutions ? NO

Corollary: SDP feasibility

- $\in \text{NP} \cap \text{co-NP}$  in real number model.
- is not NP complete in Turing model, unless  $\text{NP} = \text{co-NP}$
- After 30 years, still the best result we have.
- Very popular in computer science community.

## Part 4: Ramana's alternative system (Ram-Alt-P)

(P) is infeasible  $\Leftrightarrow \exists 0 \leq k \leq n - 1 :$

$$\begin{array}{l}
 U_0 = V_0 = 0 \\
 \left. \begin{array}{l}
 \mathcal{A}^* y_i = U_i + V_i \\
 b^T y_i = 0 \\
 U_i \in \mathbb{S}_+^n \\
 V_i \in \tan(U_{i-1})
 \end{array} \right\} \text{for } i \in [k] \\
 \left. \begin{array}{l}
 \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\
 b^T y = -1
 \end{array} \right\} \text{Like (Alt-P)}
 \end{array}$$

is feasible.

- $\tan(U)$  is tangent space of  $\mathbb{S}_+^n$  at  $U$ ,  
 $\tan(U) = \tan(U, \mathbb{S}_+^n) = \left\{ V \in \mathbb{S}^n : \text{dist}(U \pm \epsilon V, \mathbb{S}_+^n) = o(\epsilon) \right\}$
- Key point:  $\tan(U)$  is representable by SDP

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(P) is infeasible  $\Leftrightarrow \exists 0 \leq k \leq n - 1 :$

$$\begin{array}{l}
 U_0 = V_0 = 0 \\
 \mathcal{A}^* y_i = U_i + V_i \\
 b^T y_i = 0 \\
 U_i \in \mathbb{S}_+^n \\
 V_i \in \tan(U_{i-1})
 \end{array}
 \left. \vphantom{\begin{array}{l} U_0 = V_0 = 0 \\ \mathcal{A}^* y_i = U_i + V_i \\ b^T y_i = 0 \\ U_i \in \mathbb{S}_+^n \\ V_i \in \tan(U_{i-1}) \end{array}} \right\} \text{for } i \in [k]$$
  

$$\begin{array}{l}
 \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\
 b^T y = -1
 \end{array}
 \left. \vphantom{\begin{array}{l} \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\ b^T y = -1 \end{array}} \right\} \text{Like (Alt-P)}$$

is feasible.

- Naturally, at least as strong as classical alternative (Alt-P) since  $0 \in \tan(U) \forall U \succeq 0$

## Part 4: Ramana's alternative system (Ram-Alt-P)

(P) is infeasible  $\Leftrightarrow \exists 0 \leq k \leq n - 1 :$

$$\begin{array}{l}
 U_0 = V_0 = 0 \\
 \mathcal{A}^* y_i = U_i + V_i \\
 b^T y_i = 0 \\
 U_i \in \mathbb{S}_+^n \\
 V_i \in \tan(U_{i-1})
 \end{array}
 \left. \vphantom{\begin{array}{l} U_0 = V_0 = 0 \\ \mathcal{A}^* y_i = U_i + V_i \\ b^T y_i = 0 \\ U_i \in \mathbb{S}_+^n \\ V_i \in \tan(U_{i-1}) \end{array}} \right\} \text{for } i \in [k]$$
  

$$\left. \begin{array}{l}
 \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\
 b^T y = -1
 \end{array} \right\} \text{Like (Alt-P)}$$

is feasible.

- In words: in (Ram-Alt-P) we decompose each  $\mathcal{A}^* y_i$  and  $\mathcal{A}^* y$  into a psd part and a tangent space part.
- $\Leftarrow$  is easy
- $\Rightarrow$  is hard

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$$\left. \begin{array}{l} \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\ b^T y = -1 \end{array} \right\} \text{Like (Alt-P)}$$

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- Many extra variables, but very powerful properties.

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is feasible.

- Many extra variables, but very powerful properties.
- We show: proof of  $\Rightarrow$  naturally fits into reformulation framework.

## Proof sketch of $\Rightarrow$ :

$$\begin{array}{l} U_0 = V_0 = 0 \\ \mathcal{A}^* y_i = U_i + V_i \\ b^T y_i = 0 \\ U_i \in \mathbb{S}_+^n \\ V_i \in \tan(U_{i-1}) \end{array} \left. \vphantom{\begin{array}{l} U_0 = V_0 = 0 \\ \mathcal{A}^* y_i = U_i + V_i \\ b^T y_i = 0 \\ U_i \in \mathbb{S}_+^n \\ V_i \in \tan(U_{i-1}) \end{array}} \right\} \text{for } i \in [k]$$
$$\begin{array}{l} \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\ b^T y = -1 \end{array} \left. \vphantom{\begin{array}{l} \mathcal{A}^* y \in \mathbb{S}_+^n + \tan(U_k) \\ b^T y = -1 \end{array}} \right\} \text{Like (Alt-P)}$$

Key observation: **(Ram-Alt-P)** invariant under reformulating **(P)**. (E.g. rotations preserve tangent space)

So we assume infeasible **(P)** was reformulated into **(P<sub>ref</sub>)**.

## Useful fact about tangent spaces

$$\tan \begin{pmatrix} I & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} \times & \times \\ \times & 0 \end{pmatrix}$$

E.g.

$$\tan \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \ni \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Proof of  $(P)$  infeasible  $\Rightarrow$  (Ram - Alt - P) is feasible

$$\begin{array}{c}
 \underbrace{\begin{pmatrix} I_{r_1} & 0 \\ 0 & 0 \end{pmatrix}}_{A_1} \bullet X = 0 \\
 \underbrace{\begin{pmatrix} \times_{r_1} & \times & \times \\ \times & I_{r_2} & 0 \\ \times & 0 & 0 \end{pmatrix}}_{A_2} \bullet X = 0 \\
 \vdots \\
 \underbrace{\begin{pmatrix} \times_{r_1+\dots+r_k} & \times & \times \\ \times & I_{r_{k+1}} & 0 \\ \times & 0 & 0 \end{pmatrix}}_{A_{k+1}} \bullet X = -1 \\
 \vdots \\
 X \succeq 0
 \end{array} \quad (P_{\text{ref}})$$

## Matrices naturally decompose

$$\underbrace{\begin{pmatrix} \mathbf{I}_{r_1} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{pmatrix}}_{A_1 = U_1 \succeq 0}$$

$$\underbrace{\begin{pmatrix} \times_{r_1} & \times & \times \\ \times & \mathbf{I}_{r_2} & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix}}_{A_2} = \underbrace{\begin{pmatrix} \mathbf{I}_{r_1} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_{r_2} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{pmatrix}}_{U_2 \succeq 0} + \underbrace{\begin{pmatrix} \times_{r_1} & \times & \times \\ \times & \mathbf{0} & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix}}_{V_2 \in \tan(U_1)}$$

$$\underbrace{\begin{pmatrix} \times_{r_1+r_2} & \times & \times \\ \times & \mathbf{I}_{r_3} & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix}}_{A_3} = \underbrace{\begin{pmatrix} \mathbf{I}_{r_1+r_2} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_{r_3} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} \end{pmatrix}}_{U_3 \succeq 0} + \underbrace{\begin{pmatrix} \times_{r_1+r_2} & \times & \times \\ \times & \mathbf{0} & \mathbf{0} \\ \times & \mathbf{0} & \mathbf{0} \end{pmatrix}}_{V_3 \in \tan(U_2)} \dots$$

$\Rightarrow y_1 := e_1, \dots, y_k := e_k, y := e_{k+1}$  feasible in (Ram-Alt-P)

## Machinery used

- At start: subdifferentials, relative interiors, faces, conjugate faces, gauge functions, quadratic modules . . .
- More recently, only 2 ingredients:
  - Strong duality under Slater condition
  - Basic linear algebra
- Analogy with LP: in LP we need strong duality+ basic linear algebra.

## Papers

- **P**: Bad semidefinite programs: they all look the same, **2010–SIOPT 2017**
- **Liu–P**: Exact duality in semidefinite programming based on elementary reformulations, **SIOPT 2015**
- **Liu–P**: Exact duals and short certificates of infeasibility and weak infeasibility in conic linear programming, **MPA 2017**.
- **P**: Characterizing bad semidefinite programs: normal forms and short proofs, **SIREV 2017**
- **Zhu–P–Tran-Dinh**: Sieve-SDP: a simple facial reduction algorithm to preprocess semidefinite programs, **MPC, 2018**
- **Lourenco-Muramatsu-Tsuchiya**: A structural geometrical analysis of weakly infeasible SDPs, **J. Oper. Res. Soc. Jpn, 2016**
- **Permenter-Parrilo**: Partial facial reduction: simplified, equivalent SDPs via approximations of the PSD cone, **MPA, 2017**
- **Permenter-Friberg-Andersen**: Solving conic optimization problems via self-dual embedding and facial reduction **SIOPT 2017**
- **P**: A Combinatorial Approach to Ramana’s Exact Dual for Semidefinite Programming, **2024**

## On Ramana's dual

- **Ramana, Tuncel, Wolkowicz, 1997, P 2000** other correctness proof, connection to FR
- **de Klerk, Terlaky, Roos 2000** Use of Ramana's dual in self-dual embeddings.
- **P 2013** again, connection to FR, and generalization to conic LPs over **nice cones**
- **Klep, Schweighofer 2013** exact dual, with analogous properties, based on algebraic geometry
- **Liu-P 2017** Ramana dual for arbitrary conic LPs
- **Lourenço-P 2022** One more correctness proof

Thank you!