

# Strong Duality and Stability in Conic Convex Optimization

Henry Wolkowicz  
University of Waterloo

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# Cone Optimization

## Primal-Dual Pair of Optimization Problems in Conic Form

$$(finite) \quad v_P = \sup_y \{ \langle b, y \rangle : \mathcal{A}^* y \preceq_K c \}, \quad (\mathcal{P})$$

$$v_D = \inf_x \{ \langle c, x \rangle : \mathcal{A}x = b, x \succeq_{K^*} 0 \}. \quad (\mathcal{D})$$

where

- $\mathcal{A}$  - an onto linear transformation; adjoint is  $\mathcal{A}^*$
- $K$  - a proper convex cone with dual/polar cone  $K^* = \{x : \langle x, z \rangle \geq 0, \forall z \in K\}$ .
- $z' \preceq_K z'' (z' \prec_K z'')$  - partial order,  $z'' - z' \in K (\in \text{int}K)$

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# Semidefinite Programming, SDP

## SDP

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$$v_D = \inf_x \{ \langle c, x \rangle : \mathcal{A}x = b, x \succeq_{K^*} 0 \}. \quad (\mathcal{D})$$

For SDP,  $\mathcal{A} : \mathbb{S}^n \rightarrow \mathbb{R}^m$ ,  $b \in \mathbb{R}^m$ ,  $c \in \mathbb{S}^n$ , and  $K = K^* = \mathbb{S}_+^n := \{X \in \mathbb{S}^n : X \text{ is PSD}\}$ .

# Faces of Cones

## Face

A convex cone  $F$  is a **face** of  $K$ , denoted  $F \trianglelefteq K$ , if

$$x, y \in K \text{ and } x + y \in F \implies x, y \in F.$$

If  $F \trianglelefteq K$  and  $F \neq K$ , write  $F \triangleleft K$ .

## Conjugate Face

If  $F \trianglelefteq K$ , the **conjugate face** (or complementary face) of  $F$  is

$$F^c := F^\perp \cap K^* \trianglelefteq K^*.$$

If  $x \in \text{ri}(F)$ , then  $F^c = \{x\}^\perp \cap K^*$ .

## Minimal Face (Minimal Cone)

### Feasible set of $(\mathcal{P})$

Let  $\mathcal{F}_{\mathcal{P}} := \{y : c - \mathcal{A}^*y \succeq_K 0\}$

### Minimal Face

Assuming that  $\mathcal{F}_{\mathcal{P}}$  is nonempty, the **minimal face** (or minimal cone) of  $(\mathcal{P})$  is

$$f_{\mathcal{P}} := \bigcap \{F \trianglelefteq K : c - \mathcal{A}^*(\mathcal{F}_{\mathcal{P}}) \subset F\}.$$

i.e., the minimal face that contains all the feasible slacks.

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# SDP Example from Ramana, 1995

## Primal SDP

$$0 = v_P = \sup_y \left\{ y_2 : \begin{pmatrix} y_2 & 0 & 0 \\ 0 & y_1 & y_2 \\ 0 & y_2 & 0 \end{pmatrix} \succeq \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \right\}$$

$$y^* = (y_1^* \ 0)^T, \quad y_1^* \leq 0, \quad Z^* = c - \mathcal{A}^* y^* = \begin{pmatrix} 1 & 0 & 0 \\ 0 & -y_1^* & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Slater's CQ fails for primal

# Dual of SDP Example

## Dual Program

$$1 = v_D = \inf_X \{X_{11} : X_{22} = 0, X_{11} + 2X_{23} = 1, X \succeq 0\}$$

$$X^* = \begin{pmatrix} 1 & 0 & X_{13} \\ 0 & 0 & 0 \\ X_{13} & 0 & X_{33} \end{pmatrix}, \quad X_{33} \geq (X_{13}^2)$$

Slater's CQ for (primal) dual & complementarity **fails**

duality gap  $v_D - v_P = 1 - 0 = 1$ ,

$$\text{trace } X^* Z^* = \text{trace} \begin{pmatrix} 1 & 0 & X_{13} \\ 0 & 0 & 0 \\ X_{13} & 0 & X_{33} \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & -y_1^* & 0 \\ 0 & 0 & 0 \end{pmatrix} = 1 > 0$$

## Minimal Face for Ramana Example

### Feasible Set/Minimal Face

$$\mathcal{F}_P = \{y \in \mathbb{R}^2 : y_1 \leq 0, y_2 = 0\}$$

$$\begin{aligned} f_P &= \bigcap \{F \trianglelefteq K : c - \mathcal{A}^*(\mathcal{F}_P) \subset F\} \\ &= \begin{pmatrix} \mathbb{S}_+^2 & 0 \\ 0 & 0 \end{pmatrix} \\ &\triangleleft \mathbb{S}_+^3 \end{aligned}$$

### Slater CQ and Minimal Face

If  $(P)$  is feasible, then

$$c - \mathcal{A}^*y \not\prec_K 0 \quad \forall y \quad (\text{Slater's CQ fails for } (P)) \iff f_P \triangleleft K$$

## Regularization of $(\mathcal{P})$

### Borwein-W (1981)

If  $v_{\mathcal{P}}$  is finite, then  $(\mathcal{P})$  is equivalent to **regularized**  $(\mathcal{P})$

$$v_{RP} = \sup_y \{ \langle b, y \rangle : \mathcal{A}^* y \preceq_{f_P} c \}. \quad (\text{RP})$$

### Lagrangian Dual DRP Satisfies Strong Duality:

$$v_{\mathcal{P}} = v_{RP} = v_{DRP} = \inf_x \{ \langle c, x \rangle : \mathcal{A}x = b, x \succeq_{f_P^*} 0 \} \quad (\text{DRP})$$

and  $v_{DRP}$  is attained

# Implementation Problems with Regularization

## Difficulties

Borwein and W. also gave an algorithm to compute  $f_P$ .

But Difficulties:

- 1 The algorithm requires the solution of several (homogeneous) cone programs (constraints are:  
 $Ax = 0, \langle c, x \rangle = 0, 0 \neq x \succeq_K 0$ )
- 2 If Slater's CQ fails for  $(\mathcal{D})$ , then it also fails for each of these cone programs.

# Ramana's Strong Dual for SDP

Ramana '95: **Extended Lagrange-Slater dual (ELSD)** for  $(\mathcal{P})$

Construction of this dual takes advantage of the well understood facial structure of  $\mathbb{S}_+^n$ .

## Advantages:

- 1 ELSD is explicit in terms of original data  $(\mathcal{A}, b, c)$
- 2 ELSD is **poly. size** (# vrbles is  $(kn^2)$ ,  $k \leq \min\{m, n\}$ )

## Disdvantages:

- 1 Slater's CQ may fail for ELSD and its Lagrangian dual.
- 2 ELSD can potentially be very large.

## Our Goals:

### Equivalence in the case of SDP

Ramana, Tunçel, and W. '97: **Ramana's ELSD is equivalent to (DRP)** (dual of regularized primal of Borwein and W.)  
(Both approaches may require solution of potentially large SDPs that need not satisfy Slater's CQ.)

### Goals: Derive an Algorithm that Satisfies

- 1 recognizes if Slater's CQ holds and if  $(\mathcal{P})-(\mathcal{D})$  has a zero duality gap
- 2 size of any intermediate cone program solved does not exceed that of  $(\mathcal{P})$  or  $(\mathcal{D})$
- 3 intermediate cone programs to be solved are well behaved

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## Theorem of the Alternative for Slater's CQ

### THEOREM

Suppose that  $(\mathcal{P})$  is feasible. Then exactly one of the following two systems is consistent:

- (1)  $Ax = 0$ ,  $\langle c, x \rangle = 0$ , and  $0 \neq x \succeq_{K^*} 0$
- (2)  $A^*y \prec_K c$  (Slater's CQ holds for  $(\mathcal{P})$ )

### Difficult?

In theory, we can solve  $\min\{0 : x \text{ satisfies (1)}\}$  to determine if Slater's CQ fails for  $(\mathcal{P})$ .

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## Stable Theorem of the Alternative

### Stable Auxiliary Problem

Let  $e \in \text{int}(K) \cap \text{int}(K^*)$ ; define  $\mathcal{A}_c x := \begin{pmatrix} \mathcal{A}x \\ \langle c, x \rangle \end{pmatrix}$

$$\alpha^* := \left\{ \inf_{x, \alpha} \alpha : \mathcal{A}_c x = 0, x + \alpha e \succeq_{K^*} 0, \langle e, x \rangle \leq 1 \right\} \quad (\mathcal{A})$$

### Properties/Advantages

- size of  $(\mathcal{A})$  essentially that of  $(\mathcal{D})$
- A strictly feasible primal-dual point is easily found.
- Apply primal-dual IPM; assume a barrier for  $K^*$  such that the central path defined by it converges to a point in the relative interior of the optimal face; follow central path closely at end of algorithm.

# Slater's Condition and the Auxiliary problem

Solution to  $(\mathcal{A})$  yields info on  $(\mathcal{P})-(\mathcal{D})$

**Theorem:** The  $x$  component of the central path for  $(\mathcal{A})$  converges to a point in  $\text{ri}(\text{face}(G_P))$ , where

$$G_P := \{x : Ax = 0, \langle c, x \rangle = 0, x \succeq_{K^*} 0\}.$$

Moreover, since  $f_P \subset \{x^*\}^\perp \cap K = [\text{face}(G_P)]^c \trianglelefteq K$ , one of the following holds:

- 1  $\alpha^* = 0$  and  $x^* = 0$ , so Slater's CQ holds for  $(\mathcal{P})$ , or
- 2  $\alpha^* = 0$  and  $0 \neq x^* \succeq_{K^*} 0$ , so  $f_P \subset \{x^*\}^\perp \cap K \triangleleft K$ , or
- 3  $\alpha^* < 0$  and  $x^* \succ_{K^*} 0$ , so the generalized Slater CQ holds for  $(\mathcal{D})$ .

# Our Algorithm

**Input:**  $\mathcal{A}, b, c, K$ , and  $\varepsilon > 0$

Compute an optimal  $\alpha^*$  and  $x^* \in \text{ri}(\text{face}(G_P))$  from  $(\mathcal{A})$  using a primal-dual IPM.

**While**  $\|x^*\| > \varepsilon$

*If  $\alpha^* < 0$ , then  $x^* \succ_{K^*} 0$ ,  $f_P = \{0\}$ . Hence optimal  $y$  for  $(\mathcal{P})$  satisfies  $A^*y = c$ ; exit algorithm.*

**Else**

- ①  $y \in \mathcal{F}_P$  implies  $c - A^*y \in [\text{face}(G_P)]^c \triangleleft K$ , get *reduced primal* with cone  $K' = [\text{face}(G_P)]^c$ .
- ② Replace (update) primal by reduced primal.

**End**

## Conclusion of Algorithm

### Finish

Finally, solve reduced primal problem for which Slater's CQ holds.

(This provides a **certificate of optimality**.)

→ For cones such as  $\mathbb{S}_+^n$ , auxiliary problems get progressively smaller.

## Regularization for SDP

The (Conjugate) Faces,  $\mathcal{F} \trianglelefteq \mathbb{S}_+^n$  are of the Form

$$\mathcal{F} = (P \ Q) \begin{pmatrix} \mathbb{S}_+^r & 0 \\ 0 & 0 \end{pmatrix} (P \ Q)^T = P \mathbb{S}_+^r P^T$$

$$\mathcal{F}^c = (P \ Q) \begin{pmatrix} 0 & 0 \\ 0 & \mathbb{S}_+^{n-r} \end{pmatrix} (P \ Q)^T = Q \mathbb{S}_+^{n-r} Q^T$$

where matrix  $(P \ Q)$  is orthogonal.

## The Minimal Face $f_P$ Using the Auxiliary Problem

With  $x^* \in \text{ri}(G_P)$  from Auxiliary Problem

$$x^* = \begin{pmatrix} P & Q \end{pmatrix} \begin{pmatrix} D & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} P & Q \end{pmatrix}^T,$$

$$\text{face}(G_P) = \text{face}(x^*) = \begin{pmatrix} P & Q \end{pmatrix} \begin{pmatrix} \mathbb{S}_+^r & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} P & Q \end{pmatrix}^T$$

$$\text{Then: } f_P \trianglelefteq K' := [\text{face}(G_P)]^c = Q\mathbb{S}_+^{n-r}Q^T.$$

WLOG: shift  $c$  and find linear transformation  $\mathcal{L}'$

$$c' \leftarrow c - A^*y' \in K' - K'; \quad \mathcal{R}(A^*\mathcal{L}') \subset K' - K'$$

A Reduced Lower Dimensional Primal Problem

equivalent cone constraints:  $A^*\mathcal{L}'y' \preceq_{K'} c'$

$$Q^T(A^*\mathcal{L}'y')Q \preceq_{\mathbb{S}_+^{n-r}} Q^Tc'Q$$



## Previous SDP with $K = \mathbb{S}_+^3$ and a Duality Gap of 1

### SeDuMi 1.1 Results

$$y^* = \begin{pmatrix} -0.321 \times 10^6 & 0.372 \end{pmatrix}^T$$
$$s^* = \begin{pmatrix} 0.628 \times 10^5 & 0 & 0 \\ 0 & -0.321 \times 10^6 & -0.372 \\ 0 & -0.372 & 0 \end{pmatrix};$$

desired accuracy ( $10^{-6}$ ) achieved but!!

$\langle c, x^* \rangle - \langle b, y^* \rangle \approx -0.12!$  and  $s^*$  is **not** pos. semidef.

### After One Step of the Reduction

Our code yields correct primal solution:

$$y^* = \begin{pmatrix} -1.50 \\ 0 \end{pmatrix}, \quad s^* = \begin{pmatrix} 1.00 & 0 & 0 \\ 0 & 1.50 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

## Higher Dimensional Numerical Experiments

SDP with  $m = n \geq 3$ ,  $b = e_2$ ,  $c = 0$

$$\mathcal{A}^* y = \begin{pmatrix} y_1 & y_2 & y_3 & \cdots & y_{n-1} & y_n \\ y_2 & y_3 & & & & \\ \vdots & & & \ddots & & \\ y_{n-1} & & & & y_n & \\ y_n & & & & & 0 \end{pmatrix}$$

### SeDuMi/Our Algorithm

SeDuMi gives incorrect primal/dual solution; duality gap of  $-1$ ;  
our algorithm gives correct solution

$$\mathcal{F}_P = \{y \in \mathbb{R}^n : y_1 \leq 0, y_2 = \cdots = y_n = 0\}$$

$$\text{min. face } \mathcal{F}_P = \{Z \in \mathbb{S}_+^n : Z_{11} \geq 0, Z_{ij} = 0 \forall (i,j) \neq (1,1)\},$$

and  $(\mathcal{D})$  is infeasible.

**(P)–(D) in Symmetric Subspace Form****Symmetric Subspace Form**

Let:  $\bar{s} := c$ ;  $A\bar{x} = b$ ;  $\mathcal{L} = \text{Nullspace}(A)$ . Then:

$$v_P = \langle \bar{s}, \bar{x} \rangle - \inf_s \{ \langle \bar{x}, s \rangle : s \in (\bar{s} + \mathcal{L}^\perp) \cap K \}, \quad (P')$$

$$v_D = \inf_x \{ \langle \bar{s}, x \rangle : x \in (\bar{x} + \mathcal{L}) \cap K^* \}. \quad (D')$$

**Recession Cone Feasibility Problems for (P') and (D'):**

$e \in \text{int}(K) \cap \text{int}(K^*)$ ; and  $(0 \neq x^*, 0 = \alpha^*)$  soln to aux. prob.

$$\mathcal{S} := \{ s \in \mathcal{L}^\perp \cap K : \langle e, s \rangle = 1 \}, \quad (\text{PRF})$$

$$x^* / \langle e, x^* \rangle \in \mathcal{X} := \{ x \in \mathcal{L} \cap K^* : \langle e, x \rangle = 1 \}. \quad (\text{DRF})$$

# Complementarity Partition

## Symmetric Subspace Form

$$\mathcal{S} := \{s \in \mathcal{L}^\perp \cap K : \langle e, s \rangle = 1\}, \quad \mathcal{L}^\perp = \mathcal{R}(\mathcal{A}^*) \quad (\text{PRF})$$

$$\mathcal{X} := \{x \in \mathcal{L} \cap K^* : \langle e, x \rangle = 1\} \quad \mathcal{L} = \mathcal{N}(\mathcal{A}) \quad (\text{DRF})$$

## Complementarity Partition for given $\mathcal{F} \trianglelefteq K$ :

$(\mathcal{F}, \mathcal{F}^c)$  is a **complementarity partition** if

$$\text{face}(\mathcal{S}) \subset \mathcal{F} \text{ and } \text{face}(\mathcal{X}) \subset \mathcal{F}^c;$$

it is a **strict complementarity partition** if also

$$[\text{face}(\mathcal{S})]^c = \text{face}(\mathcal{X}) \text{ (equiv. } [\text{face}(\mathcal{S})]^c \cap [\text{face}(\mathcal{X})]^c = \{0\});$$

it is **proper** if  $\mathcal{S}$  and  $\mathcal{X}$  are both nonempty.

# Strict Complementarity and Nonzero Gaps

**Theorem:** Let  $K$  be a proper cone

(1) If (PRF)–(DRF) has a proper complementarity partition but not a strict complementarity partition, then there exists  $\bar{s}$  and  $\bar{x}$  such that (P)–(D) with data  $(\mathcal{L}, K, \bar{s}, \bar{x})$  has a finite nonzero duality gap.

(Partial Converse)

(2) If (a) (P)–(D) with data  $(\mathcal{L}, K, \bar{s}, \bar{x})$  has a finite nonzero duality gap with both optimal values attained, and (b) all feasible solutions of (P) and (D) are optimal, then (PRF)–(DRF) has a proper complementarity partition but not a strict complementarity partition.

## Generating SDP Instances with nonzero gaps

$K = \mathbb{S}_+^n$  Instance

Choose positive integers  $n, p, d$  with  $n > p + d$ . Let  $e = I_n \in \text{int}(K) \cap \text{int}(K^*)$ .

Choose subspace  $\mathcal{L}$  and Orthogonal Matrix  $Q$

$$\text{face}(\mathcal{L}^\perp \cap K) = Q \begin{pmatrix} 0 & & \\ & 0 & \\ & & \mathbb{S}_+^p \end{pmatrix} Q^T, \text{face}(\mathcal{L} \cap K^*) = Q \begin{pmatrix} \mathbb{S}_+^d & & \\ & 0 & \\ & & 0 \end{pmatrix} Q^T.$$

These faces form a **not strict** complementarity partition

Choose a nonzero  $U \in \mathbb{S}_+^{n-p-d}$

$$\bar{s} := \bar{x} := Q \begin{pmatrix} 0 & & \\ & U & \\ & & 0 \end{pmatrix} Q^T.$$

duality gap is  $\langle \bar{s}, \bar{x} \rangle = \|U\|_F^2 > 0$ .

# Conclusion

## Summary:

- presented a stable algorithm to solve (feasible) conic problems for which Slater's CQ fails;
- algorithm requires the solution of problems whose size is the same as that of the original dual; In special cases such as SDP and SOCP, these problems become progressively smaller;
- Failure of strict complementarity for the associated recession problems is closely related to the existence of instances having a finite nonzero duality gap; provides a means of generating instances for testing.

## Work in Progress

### Future:

- We intend to refine our code and test it on larger SDPs having a finite nonzero duality gap.
- Perform backward error analysis to study how rounding errors and errors in computing approximate solutions to the auxiliary problems affects the number of iterations of our algorithm.
- In particular, we want to reduce the problem when Slater's condition **almost** fails.



## Auxiliary Problem for Distance to Infeasibility

### Perturbed Auxiliary Problem

let  $\mathcal{Q}$  denote the second order cone, SOC; relax the equality constraints  $\mathcal{A}_c x = 0$  to SOC constraint  $\|\mathcal{A}_c x\|_2 \leq \delta$ .

$$\begin{aligned}
 v_P^{aux} &:= \inf_{x, \delta} && \delta \\
 \text{s.t.} &&& \begin{pmatrix} \delta \\ \mathcal{A}_c x \end{pmatrix} \succeq_{\mathcal{Q}} 0 \\
 &&& \langle x, e \rangle = 1 \\
 &&& x \succeq_{K^*} 0.
 \end{aligned}$$

Similar nice properties; and, near failure of Slater's CQ is identified.