

## Lecture 14

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Continuing w/ semidefinite programming

If  $\alpha$  is  $\infty$  or  $-\infty$ , then there is no optimal primal solution & same

for when  $\beta$  is  $\infty$  or  $-\infty$  (no optimal dual solution)

Even when  $\alpha$  &  $\beta$  are finite, there may not be an optimal solution

Example: Suppose 2-D spaces  $X$  &  $Y$

$$A = \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$\Phi(x) = \begin{bmatrix} 0 & x_{1,2} \\ x_{2,1} & 0 \end{bmatrix}$$

can show that  $\alpha = 0$  but there is no optimal solution

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Suppose that  $X \in \text{PSD}(K)$  is  
primal feasible

$$\Phi(X) = B \Rightarrow X = \begin{bmatrix} X_{1,1} & 1 \\ 1 & X_{2,2} \end{bmatrix}$$

In order for  $X$  to be PSD,

~~we need~~ we need  $X_{1,1} \geq 0$  &  $X_{2,2} \geq 0$

because  $e_1^+ X e_1 = X_{1,1}$  where  $e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$e_2^+ X e_2 = X_{2,2}$  "  $e_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

The determinant is  $X_{1,1} X_{2,2} - 1$

+ ~~we~~ we need this  $\geq 0$

$$\Rightarrow X_{1,1} \cdot X_{2,2} \geq 1$$

In order for this to be true, we

need that  $X_{1,1} > 0$  (couldn't be  
satisfied if

But this means that the objective  $X_{1,1} = 0$ )

$$\langle A, X \rangle = \text{Tr} \left\{ \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} X_{1,1} & 1 \\ 1 & X_{2,2} \end{bmatrix} \right\} = X_{1,1} < 0$$

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But we could pick

$$X_n = \begin{bmatrix} 1/n & 1 \\ 1 & n \end{bmatrix} \quad \text{for each positive integer } n$$

So  $X_n \in A$  (primal feasible set)

$$\& \langle A, X_n \rangle = -1/n$$

$$\text{so } \alpha > -1/n \quad \forall \text{ positive } n$$

$$\Rightarrow \alpha = 0 \quad (\text{supremum})$$

But there is no  $X \in A$  that achieves this.

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Every semidefinite program obeys weak duality, which is the statement that  $\alpha \leq \beta$ .

Proof:

trivial if  $A = \emptyset$  b/c then  $\alpha = -\infty$

or if  $B = \emptyset$  b/c then  $\beta = \infty$

So suppose  $A \neq \emptyset$  +  $B \neq \emptyset$

For primal feasible  $X \in A$  +  
dual feasible  $Y \in B$  we have

$$\langle A, X \rangle \leq \langle \Phi^+(Y), X \rangle \quad (\text{from dual constraint})$$

$$= \langle Y, \Phi(X) \rangle \quad (\text{def'n of adjoint})$$

$$= \langle Y, B \rangle \quad (\text{from primal constraint})$$

This holds for all feasible  $X$  +  $Y$

$$\Rightarrow \sup_X \langle A, X \rangle \leq \inf_Y \langle Y, B \rangle \Rightarrow \alpha \leq \beta$$

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Weak duality implies that every dual feasible  $Y \in B$

gives an upper bound of  $\langle Y, B \rangle$  on  $\alpha$

of every primal feasible  $X \in A$

gives a lower bound on  $\beta$

i.e.,

$$\langle A, X \rangle \leq \alpha \leq \beta \leq \langle B, Y \rangle$$

Strong duality is the condition that

$$\alpha = \beta$$

this is when semidefinite programming can be a very helpful analytical tool.

It doesn't ~~always~~ hold, as the following counter example demonstrates

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$$\text{Let } X = \mathbb{C}^3, Y = \mathbb{C}^2$$

$$\text{Set } A = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\Phi(X) = \begin{bmatrix} X_{1,1} + X_{2,3} + X_{3,2} & 0 \\ 0 & X_{2,2} \end{bmatrix}$$

$\Rightarrow$  PRIMAL is ~~to optimize~~

maximize  $-X_{1,1}$

subject to  $X_{1,1} + X_{2,3} + X_{3,2} = 1$

$X_{2,2} = 0$  &  $X \geq 0$

constraint  $X(2,2) = 0$  &  $X \geq 0$

imply that  $X_{2,3} + X_{3,2} = 0$

Why? consider all

$$\begin{bmatrix} X_{1,1} & X_{1,2} & X_{1,3} \\ X_{2,1} & X_{2,2} & X_{2,3} \\ X_{3,1} & X_{3,2} & X_{3,3} \end{bmatrix}$$

submatrices must  
be positive semidefinite  
 $\therefore$  this means

$$0 = X_{2,3} = X_{3,2} = X_{1,2} = X_{2,1}$$

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So this means that  $X_{1,1} = 1$   
& thus  $\alpha \leq -1$

can see that  $\alpha = -1$  by

picking  $X = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

What is the dual problem?

1st figure out the adjoint map

$$\text{Tr} \{ Y \Phi(X) \} = \text{Tr} \{ \Phi^+(Y) X \}$$

$$Y = \begin{bmatrix} Y_{1,1} & Y_{1,2} \\ Y_{2,1} & Y_{2,2} \end{bmatrix} \quad \Phi(X) = \begin{bmatrix} X_{1,1} + X_{2,3} + X_{3,2} & 0 \\ 0 & X_{2,2} \end{bmatrix}$$

$$\text{Tr} \{ Y \cdot \Phi(X) \} = Y_{1,1} (X_{1,1} + X_{2,3} + X_{3,2}) + Y_{2,2} X_{2,2}$$

If we pick

$$\Phi^+(Y) = \begin{bmatrix} Y_{1,1} & 0 & 0 \\ 0 & Y_{2,2} & Y_{1,1} \\ 0 & Y_{1,1} & 0 \end{bmatrix}$$

then  $\Phi^+(Y) X =$

$$\text{Tr} \left\{ \begin{bmatrix} Y_{1,1} & 0 & 0 \\ 0 & Y_{2,2} & Y_{1,1} \\ 0 & Y_{1,1} & 0 \end{bmatrix} \begin{bmatrix} X_{1,1} & X_{1,2} & X_{1,3} \\ X_{2,1} & X_{2,2} & X_{2,3} \\ X_{3,1} & X_{3,2} & X_{3,3} \end{bmatrix} \right\}$$

$$= Y_{1,1} X_{1,1} + Y_{2,2} X_{2,2} + Y_{1,1} X_{3,2} + Y_{1,1} X_{2,3}$$

match! so  $\Phi^+$  is as above

For the dual program, we have the constraint

$$\Phi^+(Y) \geq A \quad \text{which is that}$$

$$\begin{bmatrix} Y_{1,1} + 1 & 0 & 0 \\ 0 & Y_{2,2} & Y_{1,1} \\ 0 & Y_{1,1} & 0 \end{bmatrix} \geq 0$$

but this means that

$$\begin{bmatrix} Y_{2,2} & Y_{1,1} \\ Y_{1,1} & 0 \end{bmatrix} \geq 0 \quad \downarrow \text{so } Y_{1,1} = 0$$



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But the dual value is

$$(B, Y) = 0 \quad \text{so that}$$
$$\beta \geq 0$$

we get that  $\beta = 0$  by picking  
 $Y = 0$

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Strong duality: various conditions  
to check for when it holds.

Slater's conditions are one set:

1. If  $A \neq \emptyset$  &  $\exists Y \in \text{Herm}(Y)$   
such that  $\Phi^+(Y) > A$ , then  
 $\alpha = \beta$  & there exists a primal  
feasible  $X \in A$  such that  $(A, X) = \alpha$
2. Similar condition for dual  
If  $B \neq 0$  &  $\exists X \in \text{PSD}(X)$   
such that  $\Phi(X) = B$  &  $X > 0$   
then  $\alpha = \beta$  &  $\exists Y \in B$   
such that  $(B, Y) = \beta$

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Important idea for proof:  
 separating hyperplane theorem -  
 every closed convex set  $C$   
 can be separated from  $x \notin C$   
 by a hyperplane



Proof: Let  $K = \left\{ \begin{pmatrix} \Phi(x) & 0 \\ 0 & \langle A, x \rangle \end{pmatrix} : x \in \text{PSD}(2) \right\}$

state w/o  
proof

If  $\exists Y \in \text{Herm}(2)$  for  
 which  $\Phi^+(Y) > A$  then  $K$  is  
 a closed  
 set.

Let  $\epsilon > 0$ .

The matrix

$$\Phi^+(Y) \begin{bmatrix} B & 0 \\ 0 & \alpha + \epsilon \end{bmatrix} \text{ is not in } K$$

If it were, this would mean that there  
 would exist  $x \in \text{PSD}(2)$  such that

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$$\Phi(X) = B \quad \& \quad \langle A, X \rangle > \alpha$$

contradicting the optimality of  $\alpha$ .

Since  $K$  is closed & convex,

there exists a hyperplane that separates

(\*) from  $K$ . This hyperplane comes

as a matrix  $Y \in \text{Herm}(Y)$  &  
a real  $\lambda$  such that

$$\langle Y, \Phi(X) \rangle + \lambda \langle A, X \rangle > \langle Y, B \rangle + \lambda(\alpha + \epsilon)$$

(\*\*)

$$\forall X \in \text{PSD}(X)$$

By assumption,  $A$  is nonempty, so that

we can pick  $X_0 \in A$  for which

$$\Phi(X_0) = B, \text{ implying}$$

$$\lambda \langle A, X \rangle > \lambda(\alpha + \epsilon)$$

But since  $\alpha = \sup_X \langle A, X \rangle$ , this  
implies that  $\lambda < 0$

We can then divide (\*\*) by  $|\lambda|$  &  
thus we can choose  $\lambda = -1$  WLOG.

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if we get

$$\langle Y, \Phi(X) \rangle - \langle A, X \rangle > \langle Y, B \rangle - (\alpha + \epsilon)$$

$$\Leftrightarrow \langle \Phi^+(Y), X \rangle - \langle A, X \rangle > \langle Y, B \rangle - (\alpha + \epsilon)$$

$$\Rightarrow \langle \Phi^+(Y) - A, X \rangle > \langle Y, B \rangle - (\alpha + \epsilon)$$

$$\forall X \in \text{PSD}(X)$$

quantity on right is independent of  $X$ . This implies that

$\Phi^+(Y) - A$  is PSD. If it were not, one could choose  $X$  to amplify a negative eigenvalue of  $\Phi^+(Y) - A$  & violate the inequality

So this  $\Rightarrow$

$$\Phi^+(Y) \geq A$$

if the  $Y$  is thus dual feasible

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Set  $X$  in the last inequality to be 0

$$\Rightarrow \langle B, Y \rangle < \alpha + \epsilon$$

So we showed that  $\forall \epsilon > 0 \exists$   
a dual feasible  $Y$  such that

$$\langle B, Y \rangle < \alpha + \epsilon$$

$$\Rightarrow \beta = \inf_Y \langle B, Y \rangle < \alpha + \epsilon$$

Given that  $\alpha \leq \beta$  from weak duality,  
we have

$$\alpha \leq \beta < \alpha + \epsilon$$

Since  $\epsilon > 0$  is arbitrary, we get

$$\alpha = \beta$$

Strong duality!

can also conclude that

$\exists$  <sup>primal</sup> feasible  $X$  such that  $\alpha = \langle A, X \rangle$