

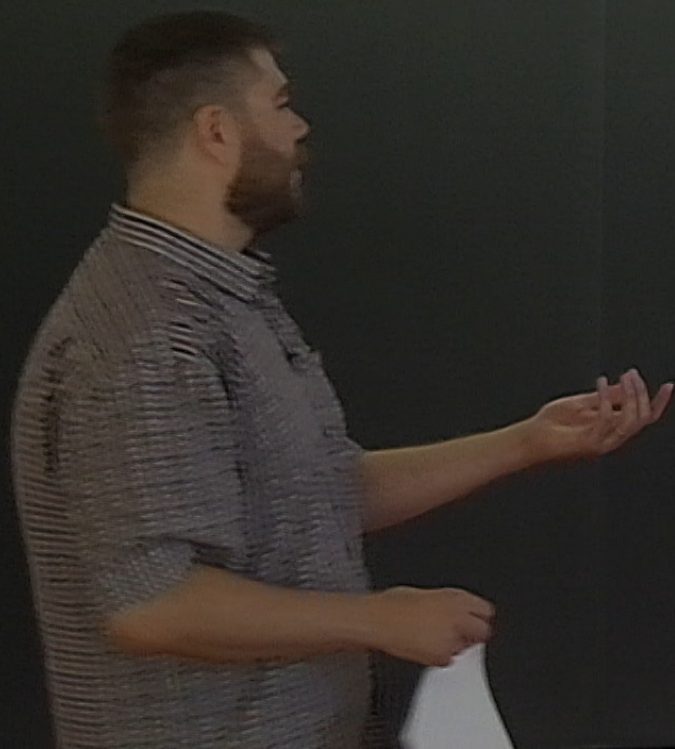
Title: SDP / Quantum Lecture Series

Speakers: Jamie Sikora

Date: July 08, 2019 - 10:00 AM

URL: <http://pirsa.org/19070069>

Introduction to SDRs



Introduction to SDPs

A Semidefinite Program (SDP)

is an optimization problem of the form

$$\alpha = \sup \langle A, X \rangle \leftarrow \text{objective function}$$

opt. value st. $\Phi(X) = B$ } constraints

$X \succeq 0$ } positive semidefinite

• $\Phi: X \rightarrow y$ linear, Hermiticity-preserving $X = X^*$
 $X = X^*$
 ≥ 0 e.vals

• $B \in \text{Herm}(y)$ $A \in \text{Herm}(X)$
 \uparrow Hermitian \uparrow finite-dim
 $X \in \mathbb{P}^k$
 $y \in \mathbb{P}^p$

• $\langle A, X \rangle = \text{Tr}(A^* X)$

X is feasible if it satisfies $\Phi(X) = B, X \succeq 0$

X is strictly feasible if it is feasible
 Φ positive definite.

Ex: $\alpha = \sup \text{Tr}(X)$ $A = I_5$
 $X = I_5$ $\Phi = 1$
 $X \succeq 0$ $B = I_5$

$X = I_5$ is the only feasible solution $X = y = \mathbb{P}^5$

$X = I_5$ is the optimal solution

$\alpha = \text{Tr}(I_5) = 5$

$$X = B, X \succeq 0$$

feasible

Ex: $\alpha = \sup \text{Tr}(X)$

$$X = -I_5$$
$$X \succeq 0$$

B from I_5 to $-I_5$

No feasible solution

SDP is infeasible

$$\alpha = -\infty$$

Ex: $\alpha = \sup \text{Tr}(X)$

$$X \succeq -I_5$$
$$X \succeq 0$$

Introduce a slack variable

$$\alpha = \sup \text{Tr}(X)$$

$$X - P = -I_5$$

$$X \succeq 0, P \succeq 0$$

$$= I_5$$
$$= 1$$
$$= I_5$$
$$y = P^5$$

$$\alpha = \sup \left\langle \begin{pmatrix} I_5 & 0 \\ 0 & 0 \end{pmatrix}, X' \right\rangle$$

$$X' = \begin{pmatrix} X & 0 \\ 0 & R \end{pmatrix}$$

$$\Phi(X') = I_5$$

$$X' \succeq 0$$

$$\Phi(X') = X - R \begin{pmatrix} \text{top-left corner} \\ \text{minus bottom-right} \end{pmatrix}$$

$X = tI_5$ for $t \geq 1$ is feasible for all $t \geq 1$.

$$\text{Tr}(X) = 5t \rightarrow +\infty \text{ as } t \rightarrow +\infty$$

SDP is unbounded, $\alpha = +\infty$

Introduction to SDPs

Finite Program (SDP)

Optimization problem of the

$p \langle A, X \rangle \leftarrow$ objective function

$\Phi(X) = B$ constraints

$X \succeq 0$ positive semidefinite

$X = X^*$
Hermiticity-preserving
 $X = X^*$
(≥ 0 e-val's)

$A \in \text{Herm}(X)$
finite-dim

Ex: $\alpha = \inf t$

$$\begin{bmatrix} t & 1 \\ 1 & s \end{bmatrix} \succeq 0$$

$t, s \in \mathbb{R}$

First, it's OK to minimize

$$\left[\begin{array}{l} -\alpha = \sup -t \\ \begin{bmatrix} t & 1 \\ 1 & s \end{bmatrix} \succeq 0 \end{array} \right]$$

$$\begin{bmatrix} t & 1 \\ 1 & s \end{bmatrix} \succeq 0 \iff t, s \geq 0, s \geq 1$$

$t > 0$

Ex: $\alpha = \sup \text{Tr}(X)$

$$X = -I_5$$

$$X \succeq 0$$

No feasible solution
SDP is infeasible.
 $\alpha = -\infty$

Ex: $\alpha = \sup \text{Tr}(X)$

$$X \succeq -I_5$$

$$X \succeq 0$$

Introduce a slack variable

$$\alpha = \sup \text{Tr}(X)$$

$$X - P = -I_5$$

$$X \succeq 0, P \succeq 0$$

$$\begin{bmatrix} t & 1 \\ 1 & s \end{bmatrix} \succeq 0$$

$$t, s \in \mathbb{R}$$

to minimize

$$\begin{bmatrix} -t & & \\ t & 1 & \\ & 1 & s \end{bmatrix} \succeq 0$$

$$\Leftrightarrow t, s \geq 0, st \geq 1$$

$$(s, t) = \left(k, \frac{1}{k}\right) \quad k > 0$$

feasible.

$\alpha = 0$ (but, no optimal solution exists)

Duality Theory

Every SDP has a dual

Primal SDP

$$\alpha = \sup \langle A, X \rangle$$

$$\begin{aligned} \Phi(X) &= B \\ X &\succeq 0 \end{aligned}$$

Adjoint: $\langle \Phi(X), Y \rangle$
 $= \langle X, \Phi^*(Y) \rangle$

Optimizing over the same data (A, B, Φ)

Dual SDP

$$\beta = \inf \langle B, Y \rangle$$

$$\Phi^*(Y) - S = A$$

$$\begin{aligned} S &\in \text{Pos}(X) \leftarrow \text{PSD} \\ Y &\in \text{Herm}(Y) \end{aligned}$$

adjoint
of Φ

Ev: $\alpha = \sup \text{Tr}(x)$
 $X = I_5$
 $X \geq 0$

$\beta = \inf \text{Tr}(y)$
 $y - S = I_5$
 $S \geq 0$
 y is "free"

$\alpha = 0$ (but, no optimal solution exists)

Duality Theory

Every SDP has a dual

Primal SDP

$$\alpha = \sup \langle A, X \rangle$$

$$\Phi(X) = B \\ X \succeq 0$$

Adjoint: $\langle \Phi(X), Y \rangle$
 $= \langle X, \Phi^*(Y) \rangle$

Optimizing over the same data (A, B, Φ)

Dual SDP

$$\beta = \inf \langle B, Y \rangle$$

$$\Phi^*(Y) - S = A \\ S \in \text{Pos}(X) \leftarrow \text{PSD} \\ Y \in \text{Herm}(Y)$$

adjoint of Φ

Ex: $\alpha = \sup \text{Tr}(X)$
 $X = I_5$
 $X \succeq 0$

$$\beta = \inf \text{Tr}(Y) \\ Y - S = I_5 \\ S \succeq 0 \\ Y \text{ is "free"}$$

$$\begin{bmatrix} t & 1 \\ 1 & s \end{bmatrix} \succeq 0$$

$t, s \in \mathbb{R}$

to minimize

$$\begin{bmatrix} t & 1 \\ 1 & s \end{bmatrix} \succeq 0$$

$\Leftrightarrow t$

$(s, t) =$
feasible

$\alpha = 0$ (but, no optimal solution exists)

Duality Theory

Every SDP has a dual

Primal SDP

$$\alpha = \sup \langle A, X \rangle$$

$$\Phi(X) = B$$

$$X \succeq 0$$

Dual SDP

$$\beta = \inf \langle B, Y \rangle$$

$$\Phi^*(Y) - S = A$$

$$S \in \text{Pos}(X) \leftarrow \text{PSD}$$

$$Y \in \text{Herm}(B)$$

Adjoint:

$$\langle \Phi(X), Y \rangle = \langle X, \Phi^*(Y) \rangle$$

using the same data (A, B, Φ)

Ex: $\alpha = \sup \text{Tr}(X)$ $\beta =$

$$X = \mathbb{I}_5$$

$$X \succeq 0$$

An optimal primal solution is $X =$

An optimal dual solution is $Y =$

$$\underline{\text{Ev:}} \quad \alpha = \sup \text{Tr}(X) \quad \beta = \inf \text{Tr}(Y)$$

$$X = I_5 \quad Y - S = I_5 \quad Y \succeq I_5$$

$$X \succeq 0 \quad S \succeq 0$$

Y is "free" Hermitian

An optimal primal solution is $X = I_5$ so $\alpha = 5$
 An optimal dual solution is $(Y, S) = (I_5, 0)$ so $\beta = 5$
 $\alpha = \beta$ is not a coincidence

Primal SDP

$$\alpha = \sup t$$

$$\begin{bmatrix} t & a & b \\ a & 0 & \frac{b}{t} \\ b & \frac{b}{t} & c \end{bmatrix} \succeq 0$$

$$\beta = \inf y$$

$$\begin{bmatrix} y & 0 & 0 \\ 0 & 0 & y \\ 0 & y & 0 \end{bmatrix} \succeq 0$$

$$\text{Ex: } \alpha = \sup \text{Tr}(X) \quad \beta = \inf \text{Tr}(Y)$$

$$X = I_5 \quad Y - S = I_5 \quad Y \succeq I_5$$

$$X \succeq 0 \quad S \succeq 0$$

Y is "free" Hermitian

An optimal primal solution is $X = I_5$ so $\alpha = 5$
 An optimal dual solution is $(Y, S) = (I_5, 0)$ so $\beta = 5$
 $\alpha = \beta$ is not a coincidence

Primal SDP

$$\alpha = \sup t$$

$$\begin{bmatrix} t & a & b \\ a & 0 & t \\ b & t & c \end{bmatrix} \succeq 0$$

$$\beta = \inf Y$$

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \succeq 0$$

If a diagonal ^{entry} on a PSD matrix is 0
that row & column are all 0's

(a, b, c, t) are feasible $\Rightarrow a = 0$

$$\frac{1-t}{2} = 0 \Rightarrow t = 1$$

$(0, 0, 0, 1)$ is feasible

$$\lambda = -1$$

For the dual: $y = 0$ in any dual feasible solution

$(y, z) = (0, 0)$ is feasible

$$\underline{\text{Ex:}} \quad \alpha = \sup_{\substack{X = \mathbb{I}_5 \\ X \succeq 0}} \text{Tr}(X) \quad \beta = \inf_{\substack{y - S = \mathbb{I}_5 \\ S \succeq 0}} \text{Tr}(y) \quad y \succeq \mathbb{I}_5$$

y is "free" Hermitian

An optimal primal solution is $X = \mathbb{I}_5$ so $\alpha = 5$

An optimal dual solution is $(y, S) = (\mathbb{I}_5, 0)$ so $\beta = 5$

$\alpha = \beta$ is not a coincidence

Ex Primal SDP

$$\alpha = \sup t$$

$$\begin{bmatrix} t & a & b \\ a & 0 & \frac{t+c}{2} \\ b & \frac{t+c}{2} & c \end{bmatrix} \succeq 0$$

$$\beta = \inf_{y, z} \begin{bmatrix} y & 0 & 0 \\ y & 0 & z \\ 0 & z & y \end{bmatrix} \succeq 0$$

matrix is 0
1 0's

> a=0

$$\frac{1-t}{2} = 0 \Rightarrow t = 1$$

any dual feasible solution
(0,0) is feasible

Weak duality

If X is feasible (i.e. $\Phi(X) = B, X \succeq 0$)
and (Y, S) is dual feasible (i.e. $\Phi^*(Y) - S = A, S \succeq 0$)

$$\text{Then } \langle A, X \rangle \leq \langle B, Y \rangle$$

In general: $\alpha \leq \beta$

Ex: $\alpha = \sup \text{Tr}(X)$
 $X = I_5$
 $X \succeq 0$

An optimal primal solution
An optimal dual solution
 $\alpha = \beta$ is not a coincidence

Ex: Primal SDP

$$\alpha = \sup t$$

$$\begin{bmatrix} t & a & b \\ a & 0 & \frac{1}{t} \\ b & \frac{1}{t} & c \end{bmatrix} \succeq 0$$

matrix is 0
1 0's

a=0

$$\frac{1-t}{z} = 0 \Rightarrow t = 1$$

any dual feasible solution
(0,0) is feasible

Weak duality

If X is feasible (i.e. $\Phi(X) = B, X \geq 0$)
and (Y, S) is dual feasible (i.e. $\Phi^*(Y) - S = A, S \geq 0$)

$$\text{then } \langle A, X \rangle \leq \langle B, Y \rangle$$

In general: $\alpha \leq \beta$

Proof: $\langle B, Y \rangle - \langle A, X \rangle$

$$= \langle \Phi(X), Y \rangle - \langle A, X \rangle$$

$$= \langle X, \Phi^*(Y) \rangle - \langle A, X \rangle$$

$$= \langle X, \Phi^*(Y) - A \rangle$$

$$= \langle X, S \rangle$$

Ex: $\alpha = \sup \text{Tr}(x)$
 $X = I_5$
 $X \geq 0$

An optimal primal solution
An optimal dual solution
 $\alpha = \beta$ is not a coincidence

Ex: Primal SDP

$$\alpha = \sup t$$

$$\begin{bmatrix} t & a & b \\ a & 0 & \frac{b}{t} \\ b & \frac{b}{t} & c \end{bmatrix} \succeq 0$$

matrix is 0
1 0's

$a=0$
 $\frac{1-t}{2}=0 \Rightarrow t=1$

any dual feasible solution
(0,0) is feasible

Weak duality

If X is feasible (i.e. $\Phi(X)=B, X \geq 0$)
and (Y,S) is dual feasible (i.e. $\Phi^*(Y)-S=A, S \geq 0$)

$$\text{then } \langle A, X \rangle \leq \langle B, Y \rangle$$

In general: $\alpha \leq \beta$

Proof:

$$\begin{aligned} \langle B, Y \rangle - \langle A, X \rangle &= \langle \Phi(X), Y \rangle - \langle A, X \rangle \\ &= \langle X, \Phi^*(Y) \rangle - \langle A, X \rangle \\ &= \langle X, \Phi^*(Y) - A \rangle \\ &= \langle X, S \rangle \\ &\geq 0 \quad \text{since } X, S \geq 0 \end{aligned}$$

An optimal primal solution
An optimal dual solution
 $\alpha = \beta$ is not a condition

Ex: Primal SDP

$$\alpha = \sup t$$

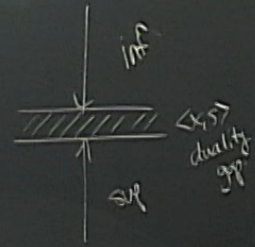
t	a	b
a	0	$\frac{1}{t} b$
b	$\frac{1}{t} a$	c

Weak duality

If X is feasible (i.e. $\Phi(X) = B, X \geq 0$)
and (Y, S) is dual feasible (i.e. $\Phi^*(Y) - S = A, S \geq 0$)

then $\langle A, X \rangle \leq \langle B, Y \rangle$

$\alpha \in I$. In general: $\alpha \leq \beta$



Proof: $\langle B, Y \rangle - \langle A, X \rangle$

$$= \langle \Phi(X), Y \rangle - \langle A, X \rangle$$

feasible solution

$$= \langle X, \Phi^*(Y) \rangle - \langle A, X \rangle$$

do

$$= \langle X, \Phi^*(Y) - A \rangle$$

$$= \langle X, S \rangle$$

$$\geq 0 \quad \text{since } X, S \geq 0$$

matrix is 0

0's

a=0

$\frac{1-t}{z}=0 \Rightarrow t=1$

Weak duality

If X is feasible (i.e. $FX=B, X \geq 0$)

and (Y, S) is dual feasible (i.e. $F^T(Y) - S = A, S \geq 0$)

then $\langle A, X \rangle \leq \langle B, Y \rangle$

In general: $\alpha \leq \beta$

Proof: $\langle B, Y \rangle - \langle A, X \rangle$

$= \langle FX, Y \rangle - \langle A, X \rangle$

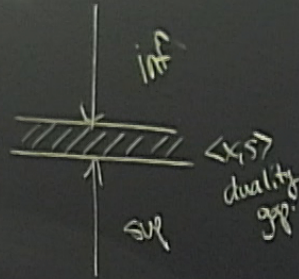
$= \langle X, F^T(Y) \rangle - \langle A, X \rangle$

$= \langle X, F^T(Y) - A \rangle$

$= \langle X, S \rangle$

≥ 0 since $X, S \geq 0$

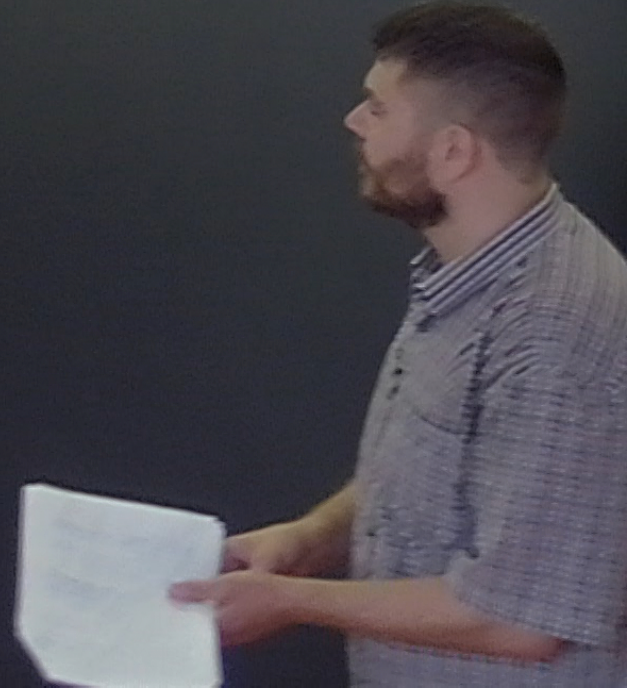
any dual feasible solution (Y, S) is feasible



Slater's Theorem: If there exists a strictly feasible solution (i.e. $F(x)=b, x>0$) and α is finite, then $\alpha^*=b$ and the dual attains an optimal solution.

Strong duality (for the primal).

x>0
duality
gap

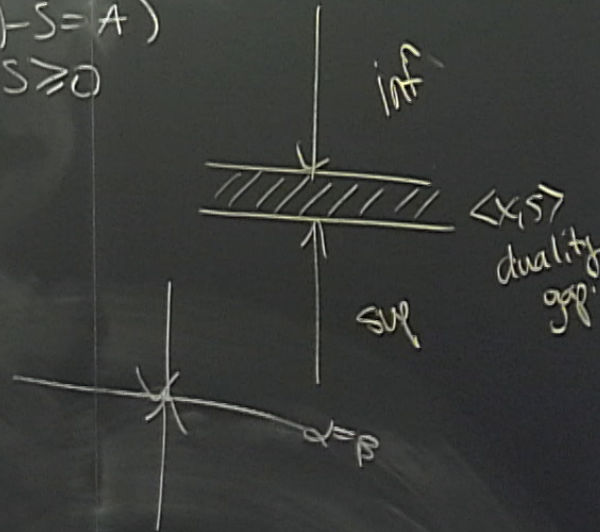


If x is feasible (i.e. $\Phi(x) = B, x \geq 0$)
 and (y, s) is dual feasible (i.e. $\Phi^*(y) - s = A$)
 $s \geq 0$

then $\langle A, x \rangle \leq \langle B, y \rangle$

I. In general: $\alpha \leq \beta$

Proof: $\langle B, y \rangle - \langle A, x \rangle$
 $= \langle \Phi(x), y \rangle - \langle A, x \rangle$
 solution $= \langle x, \Phi^*(y) \rangle - \langle A, x \rangle$
 $= \langle x, \Phi^*(y) - A \rangle$
 $= \langle x, s \rangle$
 ≥ 0 since $x, s \geq 0$



Slater's Theorem

Solution (

$\alpha = \beta$ and

Slater's Theorem: If there exists a strictly feasible solution (i.e. $F(x)=B, x>0$) and α is finite, then $\alpha=B$ and the dual attains an optimal solution.

Strong duality (for the primal)

Proof: Define the following two ^{convex} sets

$$M = \{ (S, V, t) \in (\text{Hom}(X), Q, R) : \exists X \succ S, B - F(X) = V, \langle A, X \rangle = t \}$$

$$N = \{ (0, 0, s) \in (\text{Hom}(X), Q, R) : s > \alpha \}$$

not empty as α finite.

Slater's Theorem: If there exists a strictly feasible solution (i.e. $\Phi(x) = B, x > 0$) and α is finite, then $\alpha^* = \beta$ and the dual attains an optimal solution.

Strong duality (for the primal)

Proof: Define the following two convex sets

$M = \{ (S, V, t) \in (\text{Hom}(X), Q, \mathbb{R}) : \exists X \succ S, B - \Phi(X) = V, \langle A, X \rangle = t \}$
 $N = \{ (0, 0, s) \in (\text{Hom}(X), Q, \mathbb{R}) : s > \alpha \}$

not empty as α finite.
 $Q = \{ \Phi(T) : T \in \text{Hom}(X) \} \subseteq \text{Hom}(Y)$
 \uparrow Image of Φ
 $\beta \in Q$

We defined M, N such that $M \cap N = \emptyset$.

By the separating hyperplane theorem

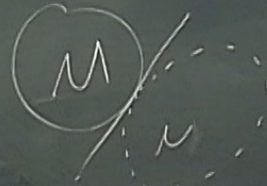
$\exists (\lambda, \gamma, \lambda) \in (\text{Hom}(X), \mathbb{Q}, \mathbb{R})$
not all 0 and $c \in \mathbb{R}$

$$(s, v, t) \in M \Rightarrow \langle \lambda, s \rangle + \langle \gamma, v \rangle + \lambda t \leq c \quad (1)$$

$$(s, v, t) \in N \Rightarrow \langle \lambda, s \rangle + \langle \gamma, v \rangle + \lambda t \geq c \quad (2)$$

(2) can be written $\lambda s \geq c \quad \forall s \geq \alpha$

$$\Rightarrow \boxed{\lambda \geq 0} + \boxed{\lambda \alpha \geq c}$$



$$M \cap N = \emptyset$$

Theorem

$$(x), Q, \mathbb{R}$$

and $C \in \mathbb{R}$

$$\langle y, v \rangle + \lambda t \leq C \quad (1)$$

$$\langle y, v \rangle + \lambda t \geq C \quad (2)$$

$$\geq C \quad \forall s > \alpha$$

$$\alpha > C$$

$$M$$
$$S \leq 0$$

Case 1: $\lambda > 0$: Rescale:

$$\lambda' = \frac{\lambda}{\alpha}, \quad y' = \frac{y}{\alpha}$$

$$(s, v, t) \in M \Rightarrow \langle \lambda', s \rangle + \langle y', v \rangle + t \leq \frac{1}{\alpha} C \leq \alpha$$

Slater's Theorem:

solution (i.e. Feasible)
 $\alpha = \beta$ and then

Proof: Define the feasible set

$$M = \{ (s, v, t) \in \text{Feasible} \}$$

$$N = \{ (0, 0, s) \in \text{Feasible} \}$$

$$M \cap N = \emptyset$$

Theorem

$$(x), Q, \mathbb{R}$$

and $C \in \mathbb{R}$

$$\langle y, v \rangle + \lambda t \leq C \quad (1)$$

$$\langle y, v \rangle + \lambda t \geq C \quad (2)$$

$$\geq C \quad \forall s \geq \alpha$$

$$\alpha \geq C$$

M
 $S \leq 0$

Case 1: $\lambda > 0$: Rescale:

$$\lambda' = \frac{\lambda}{\alpha}, \quad y' = \frac{y}{\alpha}$$

$$(s, v, t) \in M \Rightarrow \langle \lambda', s \rangle + \langle y', v \rangle + t \leq \frac{1}{\alpha} C \leq \alpha$$

Since $(s, B - \mathbb{I}(s), \langle A, s \rangle) \in M$

Set $\text{Hom}(x)$

$$\langle s, \lambda' - \mathbb{I}^*(y') + A \rangle + \langle B, y' \rangle \leq \alpha$$

Slater's Theorem:

solution (i.e. \exists)
 $\alpha = B$ and tr

Proof: Define the f

$$M = \{ (s, v, t) \in \text{Hom}(x) \mid \langle s, \lambda' - \mathbb{I}^*(y') + A \rangle + \langle B, y' \rangle \leq \alpha \}$$

$$N = \{ (0, 0, s) \in \text{Hom}(x) \mid s \geq \alpha \}$$

v such that $M \cap N = \emptyset$.

try hyperplane theorem

$A \in (\text{Hom}(X), \mathbb{Q}, \mathbb{R})$

all 0 and $C \in \mathbb{R}$

$$\langle A, s \rangle + \langle Y, v \rangle + \lambda t \leq C \quad (1)$$

$$\langle A, s \rangle + \langle Y, v \rangle + \lambda t \geq C \quad (2)$$

written $\lambda s \geq C \vee s \geq \alpha$

$$\lambda \alpha \geq C$$

$\in M$ and $S \leq 0$

$$\geq 0$$

Case 1: $\lambda > 0$: Rescale:

$$A' = \frac{A}{\lambda}, \quad Y' = \frac{Y}{\lambda}$$

$$(s, v, t) \in M \Rightarrow \langle A', s \rangle + \langle Y', v \rangle + t \leq \frac{1}{\lambda} C \leq \alpha$$

Since $(s, B^{-1}F(s), \langle A, s \rangle) \in M$

$S \in \text{Hom}(X)$

$$\langle S, A' - F^*(Y') + A \rangle + \langle B, Y' \rangle \leq \alpha$$

$$\Rightarrow A' - F^*(Y') + A = 0$$

(Y', A') is dual feasible

$$B \leq \langle B, Y' \rangle \leq \alpha \stackrel{\text{weak dual}}{=} \beta$$

$\Rightarrow \alpha = \beta$ & (Y', A') is dual optimal.

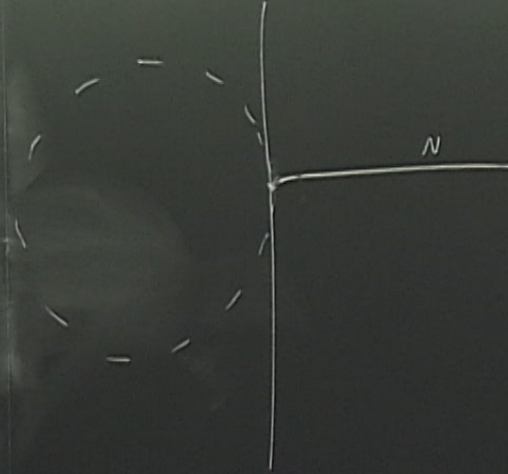
State

Proof:

$$M = \{ \dots \}$$

$$N = \{ \dots \}$$

a strictly feasible
 α is finite, then
 optimal solution
 - the primal.



conv sets
 $\exists X \geq S, B - \Phi(X) = V, \langle AX, \tau \rangle \geq t$
 α not empty as α finite.

Φ
 $Q = \{ \Phi(T) \mid T \in \text{Herm}(X) \} \subseteq \text{Herm}(y)$

Case: $\lambda = 0$

Argue that $\lambda = \gamma = 0$ as well.

• Since $(x, 0, \langle Ax \rangle) \in M$ for strictly feasible x

$$\langle A, x \rangle \leq c \leq \lambda \alpha = 0$$

But, $x > 0, \lambda \geq 0$

$$\Rightarrow \boxed{\lambda = 0}$$

$$\langle x, \lambda \rangle \geq 0 \text{ if } x, \lambda \geq 0$$

$$\langle x, \lambda \rangle > 0 \text{ if } x > 0, \lambda \neq 0$$

$$(S, v, t) \in M \Rightarrow$$

$$(S, v, t) \in N \Rightarrow \langle A, S \rangle$$

(2) can be

$$\Rightarrow \boxed{\lambda}$$

Notice for any

$$\Rightarrow$$

Since $(S, B - \Phi(S), \langle A, S \rangle) \in M$
 $S \in \text{Hom}(X)$

$$\langle \underbrace{y}_Q, \underbrace{B - \underbrace{\Phi(S)}_Q} \rangle \leq c \leq 0$$

$$\langle y, B \rangle \leq \langle y, \Phi(S) \rangle$$

\downarrow
 $y=0$

well.

$$\alpha = 0$$

Case 1: $\lambda > 0$: Rescale:

$$\lambda' = \frac{\lambda}{\lambda}, \quad y' = \frac{y}{\lambda}$$

$$(S, v) \in M \Rightarrow \langle \lambda' S, \lambda' v \rangle + c \leq \frac{1}{\lambda} c$$

Since $(S, B - \Phi(S), \langle A, S \rangle) \in M$
 $S \in \text{Hom}(X)$

$$\langle S, \lambda' - \Phi'(y') + A \rangle + \langle B, y' \rangle$$

$$\Rightarrow \lambda' - \Phi'(y') + A = 0$$

(y', λ') is dual feasible

$$B \leq \langle B, y' \rangle \leq \alpha \leq \beta$$

\downarrow
weak dual

$\Rightarrow \alpha = \beta$ & (y', λ') is dual optimal.

Since $(S, B - \Phi(S), \langle A, S \rangle) \in M$
 $S \in \text{Hom}(X)$

$$\langle \underbrace{Y}_Q, \underbrace{B - \Phi(S)}_Q \rangle \leq C \leq 0$$

$$\langle Y, B \rangle \leq \langle Y, \Phi(S) \rangle$$

\downarrow
 $y=0$

Contradiction

□

Case 1: $\lambda > 0$: Rescale:

$$\lambda' = \frac{\lambda}{\lambda}, \quad y' = \frac{y}{\lambda}$$

$$(S, v) \in M \Rightarrow \langle \lambda' S, \lambda' \rangle + \langle y', v \rangle + t \leq \frac{1}{\lambda} C \leq \alpha$$

Since $(S, B - \Phi(S), \langle A, S \rangle) \in M$
 $S \in \text{Hom}(X)$

$$\langle S, \lambda' - \Phi'(y') + A, \lambda' \rangle + \langle B, y' \rangle \leq \alpha$$

$$\Rightarrow \lambda' - \Phi'(y') + A = 0$$

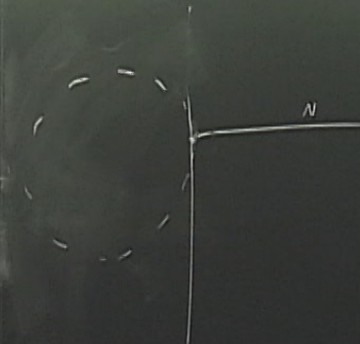
(y', λ') is dual feasible

$$B \leq \langle B, y' \rangle \leq \alpha \stackrel{\text{weak duality}}{=} \beta$$

$\Rightarrow \alpha = \beta$ & (y', λ') is dual optimal.

Slater's Theorem: If there exists a strictly feasible solution (i.e. $\exists x \succ 0, Ax = b$) and α is finite, then $\alpha^* = \beta$ and the dual attains an optimal solution.

Strong duality (for the primal)



Proof: Define the following two convex sets

$$M = \{ (S, V, t) \in \text{Hom}(X), Q, \mathbb{R} \} : \exists x \succ 0, B = \Phi(x), Ax \succ t \}$$

$$N = \{ (0, 0, s) \in \text{Hom}(X), Q, \mathbb{R} \} : s > \alpha \}$$

\uparrow Image of Φ not

$Q = \{ \Phi(x) \mid x \in \text{Hom}(X) \} = \text{Hom}(Y)$

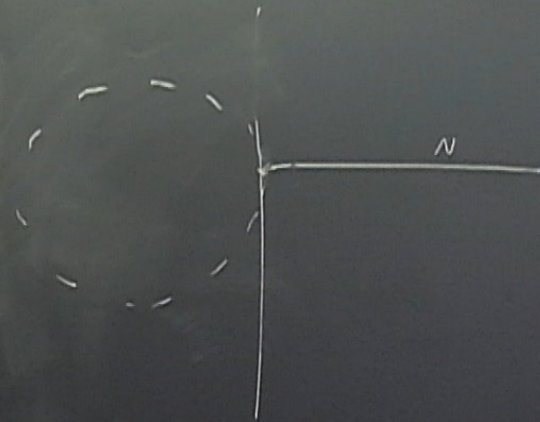
BEQ

Slater's Theorem: If there exists a strictly feasible solution (i.e. $\exists(x) = B, x > 0$) and α is finite, then $\alpha = \beta$ and the dual attains an optimal solution.

Strong duality (for the primal).

Slater for the dual: If the dual is strictly feasible (i.e. $\exists^*(y) - S = A, S > 0$) and β is finite, then $\alpha = \beta$ and α is attained.

Corollary: If both the primal and dual are strictly feasible, then $\alpha = \beta$ and both are attained.



Ex: The ground energy of a Hamiltonian H can be computed as

$$\alpha = \inf_{\substack{\langle \psi, H \rangle \leftarrow \text{energy} \\ \text{Tr}(\rho) = 1 \\ \rho \geq 0}} \langle \psi, H \rangle \quad \left. \vphantom{\inf} \right\} \text{quantum states}$$

Since $(\sum_i \rho_i \mathbb{I} - \mathbb{I}, \langle A, \rho \rangle) \in M$
 $S \in \text{Hom}(X)$

$$\langle \psi, \underbrace{\rho}_Q - \underbrace{\mathbb{I}}_Q \rangle \leq c \leq 0$$

$$\mathbb{I}(x) = \rho$$

$$\langle \psi, \rho \rangle \leq \langle \psi, \mathbb{I}(S) \rangle$$

\downarrow
 $\psi = 0$

Contradiction.

□

Case 1: λ

$$\rho \in M$$

Since $S \in \text{Hom}(X)$

$\Rightarrow \lambda'$

$$\rho \leq \lambda \mathbb{I}$$

\uparrow
defn

$\Rightarrow \alpha = \beta$ & dual optima

Ex: The ground energy of a Hamiltonian H can be computed as

$$\alpha = \inf \langle S, H \rangle \leftarrow \text{energy}$$

$$\left. \begin{array}{l} \text{Tr}(S) = 1 \\ S \geq 0 \end{array} \right\} \text{quantum states}$$

$S = \frac{1}{\dim \mathbb{1}}$ is strictly feasible.

The dual is

$$\beta = \sup \alpha$$

$$t \mathbb{1} \leq H$$

$t < \lambda_{\min}(H)$ is strictly feasible

$$S = | \psi \rangle \langle \psi | \quad \langle H, S \rangle = \lambda_{\min}(H) / N$$

$$\langle H, S \rangle = \lambda_{\min}(H)$$

Optimal t :

Case 1: λ

$\lambda' =$
 $(S, V) \in \mathcal{M}$
 $H \leq \lambda' S$
 Since
 Set $\lambda =$

$\Rightarrow \lambda' =$

$\beta \leq \lambda'$
 def'n of

$\Rightarrow \alpha = \beta$ &
 dual optimal

Ex: The ground energy of a Hamiltonian H can be computed as

$$\alpha = \inf \left\{ \langle \rho, H \rangle \leftarrow \text{energy} \right. \\ \left. \begin{array}{l} \text{Tr}(\rho) = 1 \\ \rho \geq 0 \end{array} \right\} \text{ quantum states}$$

$\rho = \frac{1}{\dim \mathbb{I}} \mathbb{I}$ is strictly feasible.

The dual is

$$\beta = \sup \alpha$$

$$\alpha \mathbb{I} \leq H$$

$\alpha < \lambda_{\min}(H)$ is strictly feasible

$$\rho = |\psi\rangle\langle\psi| \quad \langle H, \rho \rangle = \lambda_{\min}(H) \\ \langle H, \rho \rangle = \lambda_{\min}(H)$$

Optimal $\alpha = \lambda_{\min}(H)$

$$\alpha \leq \lambda_{\min}(H)$$

\forall

$$\beta \geq \lambda_{\min}(H)$$

Weak duality certifies optimality

Optimal $z = \lambda_{\min}(H)$

$$\alpha \leq \lambda_{\min}(H)$$

\forall

$$\beta \geq \lambda_{\min}(H)$$

Weak duality certifies optimality

Complementary Slackness:

$$\beta - \alpha = \langle x, s \rangle \geq 0$$

$\uparrow \uparrow$
optimal solutions

If $\alpha = \beta$ then $\langle x, s \rangle = 0$

Slater's

solut

$$\alpha = \beta$$

Slater's

(ie,

Condit

Optimal $z = \lambda_{\min}(H)$

$$\alpha \leq \lambda_{\min}(H)$$

\forall

$$\beta \geq \lambda_{\min}(H)$$

Weak duality certifies optimality

Complementary Slackness:

$$\beta - \alpha = \langle x, s \rangle \geq 0$$

\uparrow
optimal solutions

$$\text{If } \alpha = \beta \text{ then } \langle x, s \rangle = 0$$

$$\text{If } x, s \geq 0 \text{ and } \langle x, s \rangle = 0$$

$$\Rightarrow xs = 0$$

$$\text{If } \alpha = \beta + \begin{cases} SX = 0 \\ Xs = 0 \end{cases}$$

X is primal opt.
 (X, s) is dual opt.

$$X^T \bar{y} = b \quad \& \quad \bar{y}^T X = AX$$

Slater's

solut
 $\alpha = \beta$

Slater's