

Lecture 7

①

Example of an SDP

spectral norm of Hermitian operator H

Recall $\|H\|_\infty =$ largest singular value of H

$$= \max \{ |\lambda_{\max}|, |\lambda_{\min}| \}$$

Define $f(H) := \sup_{\substack{X_1, X_2 \succeq 0 \\ \text{Tr}[X_1 + X_2] \leq 1}} \{ \text{Tr}[H(X_1 - X_2)] \}$

$$\hat{f}(H) = \inf_{t \geq 0} \{ t : -tI \leq H \leq tI \}$$

Theorem: $f(H) = \hat{f}(H) = \|H\|_\infty$

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1st show that $f(H)$ can be written

as

$$S(\Phi, A, B) = \sup_{X \succeq 0} \{ \text{Tr}[AX] : \Phi(X) \in B \}$$

Pick $X = \begin{bmatrix} X_1 & z^+ \\ z & X_2 \end{bmatrix},$

$$A = \begin{bmatrix} H & 0 \\ 0 & -H \end{bmatrix},$$

$$\Phi(X) = \text{Tr}[X_1 + X_2], \quad B = 1$$

$$X \succeq 0 \Rightarrow X_1, X_2 \succeq 0$$

Notice that z plays no role in the objective nor in the constraint. ~~So~~ So it

can be eliminated & we can set

$$X = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix}$$

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recall that spectral norm is

$$\text{given by } \|H\|_2 = \max \left\{ |\lambda_{\max}|, |\lambda_{\min}| \right\}$$

Let $|\phi_{\max}\rangle$ satisfy $H|\phi_{\max}\rangle = \lambda_{\max}|\phi_{\max}\rangle$

" $|\phi_{\min}\rangle$ " $H|\phi_{\min}\rangle = \lambda_{\min}|\phi_{\min}\rangle$

• Suppose $\lambda_{\max} \geq 0$

Then choose $X_1 = |\phi_{\max}\rangle\langle\phi_{\max}|$ &
 $X_2 = 0$

$$\Rightarrow f(H) \geq \lambda_{\max}$$

• Suppose $\lambda_{\max} \leq 0$

Then choose $X_1 = 0$, $X_2 = |\phi_{\max}\rangle\langle\phi_{\max}|$

$$\Rightarrow f(H) \geq -\lambda_{\max} = |\lambda_{\max}|$$

we conclude that $f(H) \geq |\lambda_{\max}|$

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Can play a similar game
to conclude that

$$f(H) \geq |d_{\min}|$$

So we find that $f(H) \geq \max\{|d_{\max}|, |d_{\min}|\}$

Now let's show that

$$f(H) \leq \hat{f}(H)$$

To do so, we show that

$\hat{f}(H)$ is the SDP dual
of $f(H)$ & we apply
weak duality

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To find dual SDP, we need to find Φ^+ (this is always the main issue when finding the dual.)

Since $\Phi(x) = \text{tr}[X_1 + X_2]$ & $B=1$ are scalars, take $Y=t$ to be a scalar.

Then

$$\begin{aligned}\text{Tr}[Y\Phi(x)] &= t \text{Tr}[X_1 + X_2] \\ &= \text{Tr} \begin{bmatrix} tI & 0 \\ 0 & tI \end{bmatrix} \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix} \\ &= \text{Tr}[\Phi^+(Y) X]\end{aligned}$$

$$\Rightarrow \Phi^+(Y) = \Phi^+(t) = \begin{bmatrix} tI & 0 \\ 0 & tI \end{bmatrix}$$

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Now plug into standard form
of SDP dual:

$$\inf_{Y \succeq 0} \left\{ \text{Tr}[BY], \text{Tr}(Y) \geq A \right\}$$

$$= \inf_{t \geq 0} \left\{ t : \begin{bmatrix} tI & 0 \\ 0 & tI \end{bmatrix} \succeq \begin{bmatrix} H & 0 \\ 0 & -H \end{bmatrix} \right\}$$

$$= \inf_{t \geq 0} \left\{ t : tI \succeq H, tI \succeq -H \right\}$$

$$= \inf_{t \geq 0} \left\{ t : -tI \leq H \leq tI \right\}$$

$$= \hat{f}(H)$$

Consider that

$$\lambda_{\min} I \leq H \leq \lambda_{\max} I$$

$$\& \text{ also that } \lambda_{\max} I \leq \|H\|_{\infty} I$$

$$\& -\|H\|_{\infty} I \leq \lambda_{\min} I$$

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$$- \|H\|_{\infty} I \leq H \leq \|H\|_{\infty} I$$

$\Rightarrow \|H\|_{\infty}$ is a feasible choice
for t in $\hat{f}(H)$

$$\Rightarrow \hat{f}(H) \leq \|H\|_{\infty}$$

Putting everything together, we get

$$\|H\|_{\infty} \leq f(H) \leq \hat{f}(H) \leq \|H\|_{\infty}$$

\uparrow
1st part
of proof

\uparrow
weak
duality

\uparrow
2nd
part of proof

done...

(8)

Can evaluate complementary slackness conditions & they reduce to

(Recall)

$$YB = Y\Phi(X)$$
$$\Phi(Y)X = AX$$

$$t = t \operatorname{Tr}[X_1 + X_2]$$

$$\begin{bmatrix} tI & 0 \\ 0 & tI \end{bmatrix} \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix} = \begin{bmatrix} H & 0 \\ 0 & -H \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix}$$

$$\Leftrightarrow tX_1 = HX_1, \quad -tX_2 = HX_2$$

If we know eigenvalues of H are non-zero, then we conclude that

$$t \neq 0 \text{ \& } \operatorname{Tr}[X_1 + X_2] = 1$$

so it suffices to optimize over X_1, X_2 satisfying other conditions imply that

X_1, X_2 commute w/ H .

Bring up Matlab code...