

Matrix Norms

Given matrix norm $\|\cdot\|$ on $\mathbb{R}^{m \times m}$, the dual norm $\|\cdot\|_d$ is defined as

$$\|X\|_d = \sup \{ \langle X, Y \rangle : Y \in \mathbb{R}^{m \times m}, \|Y\| \leq 1 \}$$

$(\langle X, Y \rangle \equiv \text{tr } X^T Y)$

For vector p -norms: the dual of l_p norm is l_q norm, with $\frac{1}{p} + \frac{1}{q} = 1$ (Hölder's inequality) and dual of l_1 norm is l_∞ norm.

Consider $\|X\| = \|X\|_F = \langle X, X \rangle^{1/2} = (\text{tr } X^T X)^{1/2}$

Then $\|X\|_d = \|X\|_F$ (just as dual of l_2 is l_2).

How about the dual of $\|X\|_2$? (operator norm, spectral norm).
 $= \max \sigma_i(X)$

Then the dual of $\|\cdot\|_2$ is the NUCLEAR NORM

$$\|X\|_* = \sum_{i=1} \sigma_i(X)$$

(Schatten 1-norm)
("trace" norm)

To prove this we'll characterize $\|X\|_2$ and $\|X\|_*$ by SDPs.

FIRST: Characterization of $\|Z\|_2$:

$$\|Z\|_2 \leq t \Leftrightarrow t^2 I_m - Z Z^T \succeq 0 \Leftrightarrow t^2 I_m - Z^T Z \succeq 0$$

$$\Leftrightarrow \begin{bmatrix} t I_m & Z \\ Z^T & t I_m \end{bmatrix} \succeq 0 \Leftrightarrow \begin{bmatrix} t I_m & Z^T \\ Z & t I_m \end{bmatrix} \succeq 0$$

MN2

2.

Pf Use question 1 in HW

or use Schur complement (see BV p. 650):

$$\text{Let } M = \begin{bmatrix} A & B \\ B^T & C \end{bmatrix}$$

If $A > 0$, then Schur complement $S = C - B^T A^{-1} B$ ≥ 0 iff $M \geq 0$.(Block Gauss elimination:
subtract $B^T A^{-1} * 1^{st}$ row
from 2^{nd} row:

$$\begin{bmatrix} A & B \\ 0 & -B^T A^{-1} B + C \end{bmatrix}$$

Hence

$$\|Z\|_2 = \inf \left\{ t : \begin{bmatrix} tI & Z \\ Z^T & tI \end{bmatrix} \geq 0 \right\}$$

an SDP.Characterization of $\|X\|_*$.

$$\text{Let } X = U \Sigma V^T$$

$$\begin{matrix} m \\ \downarrow \end{matrix} \begin{bmatrix} \\ \\ \end{bmatrix} \begin{bmatrix} I \\ \\ \end{bmatrix} \begin{matrix} n \\ \downarrow \end{matrix}$$

$$U \quad m \times r \quad U^T U = I$$

$$V \quad n \times r \quad V^T V = I$$

$$\Sigma \quad r \times r \text{ diagonal.}$$

$$r = \text{rank}(X).$$

Then by def'n, $\|X\|_* = \text{tr } \Sigma$.

$$\text{Let } Y = UV^T. \text{ Note } \|UV^T\|_2 = \max_{\|q\|_2=1} \|UV^T q\| = 1$$

(the SVD of UV^T is UIV^T).

so

$$\|X\|_{2,d} = \sup \{ \langle X, Y \rangle : \|Y\| \leq 1 \}$$

$$\begin{aligned} &\geq \text{tr } X^T UV^T = \text{tr } V \Sigma U^T UV^T \\ &= \text{tr } \Sigma V^T V \\ &= \text{tr } \Sigma = \|X\|_*. \end{aligned}$$

MN3. To find $\|X\|_{2,d}$ we need

$$\max_Y \langle X, Y \rangle$$

s.t. $\|Y\|_2 \leq 1$

MINUS SIGN IS FOR CONVENIENCE BELOW

$$\equiv \max_Y \text{tr} X^T Y$$

Var. $Y \in \mathbb{R}^{m \times n}$
Fixed: $X \in \mathbb{R}^{m \times n}$

(D) s.t. $\begin{bmatrix} I_m & -Y \\ -Y^T & I_n \end{bmatrix} \succeq 0$

$\leftarrow \in S^{m+n}$

This is an SDP in Dual form: Let's write $B \equiv X$

(D) $\max \sum_{\substack{i=1, \dots, m \\ j=1, \dots, n}} b_{ij} y_{ij}$

s.t. $\begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} - \sum_{\substack{i=1, \dots, m \\ j=1, \dots, n}} y_{ij} \underbrace{\begin{bmatrix} & & & -1 \\ & & & \\ & & & \\ -1 & & & \end{bmatrix}}_{2E_{ij} \in S^{m+n}} \succeq 0$

(D'') i.e. $\sum_{i,j} y_{ij} E_{ij} \leq \frac{1}{2} \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$

The Primal SDP is

(P) $\min_{W \in S^{m+n}} \frac{1}{2} \left\langle \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}, \underbrace{\begin{bmatrix} W_1 & W_3 \\ W_3^T & W_2 \end{bmatrix}}_W \right\rangle$

s.t. $\langle E_{ij}, W \rangle = b_{ij}$ $i=1, \dots, m$
 $j=1, \dots, n$

$W \succeq 0 \implies (W_3)_{ij} = x_{ij}$

MN4.

This has the feasible point

$$W = \begin{bmatrix} U \Sigma U^T & U \Sigma V^T \\ V \Sigma U^T & V \Sigma V^T \end{bmatrix} = \begin{bmatrix} U \\ V \end{bmatrix} \Sigma \begin{bmatrix} U^T & V^T \end{bmatrix}$$

$$\succeq 0$$

since $W_3 = U \Sigma V^T \equiv X \equiv B$.

Also the corresponding primal objective value is

$$\frac{1}{2} (\text{tr } W_1 + \text{tr } W_2) = \text{tr } \Sigma = \|X\|_*,$$

Now any feasible point for (P) is an upper bound for the optimal solution of (D); so

$$\|X\|_{2,d} \leq \|X\|_*.$$

Combining this with $\|X\|_{2,d} \geq \|X\|_*$ (p.MN2)

we have $\|X\|_{2,d} = \|X\|_*$.

Hence by SDP duality (as (P), (D) both have strictly feasible points),

$\|X\|_* \equiv$ solution of the SDP

$$(P) \quad \min_{\substack{W_1 \in S^m \\ W_2 \in S^m}} \frac{1}{2} (\text{tr } W_1 + \text{tr } W_2)$$

$$\text{s.t. } \begin{bmatrix} W_1 & X \\ X^T & W_2 \end{bmatrix} \succeq 0.$$

MN5.

Matrix Completion The Netflix Problem.

$$\text{Given } X = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

↑ only some entries known
Believe X to be low rank.

Would like to solve

$$\begin{aligned} &\min_{X \in \mathbb{R}^{m \times n}} \text{rank}(X) \\ &\text{s.t. } X_{ij} = m_{ij} \quad (i,j) \in \Omega \\ &\quad \text{(given values)} \end{aligned}$$

NP-hard!

But, just as ℓ_1 minimization for vectors "encourages" sparsity, nuclear norm minimization for matrices "encourages" low rank — so solve

$$\begin{aligned} &\min_X \|X\|_* \\ &\text{s.t. } X_{ij} = m_{ij} \quad (i,j) \in \Omega \end{aligned}$$

ie., the SDP

$$\begin{aligned} &\min && \frac{1}{2}(\text{tr } W_1 + \text{tr } W_2) \\ &W_1 \in S^m && \\ &W_2 \in S^n && \\ &X \in \mathbb{R}^{m \times n} && \\ &\text{s.t. } && \begin{bmatrix} W_1 & X \\ X^T & W_2 \end{bmatrix} \succeq 0 \\ &&& X_{ij} = m_{ij} \quad (i,j) \in \Omega. \end{aligned}$$

MN6

More generally

$$\begin{aligned} \min_{X \in \mathbb{R}^{m \times n}} \text{rank}(X) \\ \text{s.t. } A(X) = b \\ \quad \quad \quad \leftarrow \text{LINEAR MAP.} \end{aligned}$$

Can represent as

$$\langle A_k, X \rangle = b_k \quad i=1, \dots, p$$

(like constraints of primal standard form SDP, except there A_i, X are symmetric; here they are not).

Earlier case

$$A_k = \begin{bmatrix} & & \\ & 1 & \\ & & \\ & & \\ & & \\ & & \end{bmatrix}_i$$

N.N. Relaxation:

$$\begin{aligned} \text{Primal: } \min_{X \in \mathbb{R}^{m \times n}} \quad & \frac{1}{2} (t_1 W_1 + t_2 W_2) \\ & W_1 \in S^m \\ & W_2 \in S^n \\ \text{s.t. } & \begin{bmatrix} W_1 & X \\ X^T & W_2 \end{bmatrix} \succeq 0 \\ & A(X) = b \end{aligned}$$

Dual: see HW.

$$\begin{aligned} \max \quad & b^T z \\ \text{s.t. } & \begin{bmatrix} I_m & A^*(z) \\ (A^*(z))^T & I_n \end{bmatrix} \succeq 0 \end{aligned}$$

Lecture continues at bottom of p. 478
of Recht, Fazel + Parrilo

where

$$A^*(z) = \sum_{k=1}^p z_k A_k$$