

## More on SDPs

Recall a semi-definite program (SDP) in standard form is:

$$\min_{X \in S^n} \langle C, X \rangle \leftarrow = \text{tr}(C^T X) \text{ aka } C \cdot X$$

st.  $\langle A_i, X \rangle = b_i, i=1, \dots, p$

$S^n = \text{symm. or Hermitian } n \times n \text{ matrices}$

$X \succeq 0 \leftarrow \text{aka } X \in S^n_+$

Another way to write this:

How to characterize a linear function on  $\mathbb{R}^{n \times n}$ ?

Simple: "vectorize" input  $X \mapsto \text{vec}(X) =: x$

$\underbrace{\mathbb{R}^{n \times n}} \quad \underbrace{\mathbb{R}^{n^2 \times 1}} \quad \text{lowercase}$

then let  $A$  be a  $p \times n^2$  matrix:

$$A(\text{vec}(X)) = b$$

means

$$\begin{aligned} \underbrace{\begin{bmatrix} \vdots \\ \text{row } i \\ \vdots \end{bmatrix}}_{A} \cdot \underbrace{\begin{bmatrix} x \\ \vdots \\ x \end{bmatrix}}_{\text{vec}(X)} &= \underbrace{\begin{bmatrix} b \\ \vdots \\ b \end{bmatrix}}_b \\ \left[ A(\text{vec}(X)) \right]_i &= b_i \\ &= a_i^T \cdot \text{vec}(X) \end{aligned}$$

each row of  $A$   
is  $\text{vec}(A_i)$

$$\begin{aligned} &= \langle \text{mat}(a_i), X \rangle \\ &= \langle A_i, X \rangle \end{aligned}$$

## Linear Matrix Inequality (LMI)

Closely related to SDPs (we'll see later that they are the duals of SDPs)

$$\min_{X \in \mathbb{R}^P} \langle C, X \rangle \text{ s.t. } \sum_{i=1}^P x_i \cdot A_i + G \preceq 0$$

$\uparrow$  vector  
and  $A_0 \cdot X = b$

$\uparrow \in S^n$

if  $A_i$  ( $i \in [p]$ ) and  $G$  are diagonal, this is a LP!

In a SDP, if we force  $X$  to be diagonal  
(can be enforced via linear constraints), it's a LP!

Since  $\text{diag}(x) \succeq 0$  iff  $x \succeq 0$

Can also cast SOCPs (hence QPs, QCQPs)

as SDPs using the **Schur Complement**

(c.f. problem 4.40 and § A.5.5 in Boyd & Vandenberghe)  
i.e.

$$A \in S_{++}^r, C \in S^s, B \in \mathbb{R}^{r \times s}$$

then

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

### Product cones

Often we want  $X \succeq 0$  and  $Y \succeq 0$

So can combine as  $Z = \begin{bmatrix} X & W \\ W^T & Y \end{bmatrix}$

and enforce  $Z \succeq 0, W = 0$

(mathematically OK, but computationally a bad idea.

Good software can handle multiple cone constraints

(explicitly)

ex: SeDuMi, SDPT3, Mosek, Gurobi

SDPs

## Aside (optional)

Matrix Inversion Lemma

Very useful!

aka Sherman-Morrison-Woodbury Formula

$$\underbrace{(A + UCV^T)^{-1}}_{n \times n} = A^{-1} - \underbrace{A^{-1}U}_{n \times n} \underbrace{(C^{-1} + V^T A^{-1} U)^{-1}}_{p \times p} V^T A^{-1}$$

viz Schur complement on

$$\begin{array}{c} \begin{array}{cc} n & p \\ \begin{array}{c|c} A & U \\ \hline V^T & -C^{-1} \end{array} & \begin{array}{c} n \\ p \end{array} \end{array}$$

often  $p=1$

## Aside (optional) Saddle pt. systems

Block linear systems of the form

$$\begin{bmatrix} A & B \\ B^T & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} c \\ d \end{bmatrix} \quad \text{i.e.} \quad \begin{array}{l} Ax + By = c \\ B^T x = d \end{array}$$

One way to solve:  $x = A^{-1}(c - By)$

$$\underbrace{B^T A^{-1} B}_\text{often positive definite} y = B^T A^{-1} c - d$$

often  $A$  is diagonal  
usually  $A \in \mathbb{R}^n$