

## Topics in Convex Optimization

3.1

1.  $V \subseteq \mathbb{R}^n$   $F_V = \{Y \in S_+^n : \text{im } Y \subseteq V\}$  is a face of the positive semidefinite cone

Proof:

$F_V$  is convex :  $\forall X, Y \in F_V$  ,  $(\lambda X + \mu Y)x = \lambda Xx + \mu Yx \in V \quad \forall x \in \mathbb{R}^n$

Assume  $A, B \in S_+^n$  such that  $\lambda A + \mu B \in F_V$  for some  $\lambda, \mu \in \mathbb{R}_{\geq 0}$ .

2.  $X \in S_+^n$  . Smallest closed face of  $S_+^n$  containing  $X$  is  $F_{\text{im } X}$  .

Proof:

$F_{\text{im } X}$  is a face as  $\text{im } X$  is a subspace of  $\mathbb{R}^n$ .

Any face  $F$  containing  $X$  must have  $\text{im } X \subseteq V$  by definition. Hence  $F_{\text{im } X}$  is smallest face.

$F_{\text{im } X}$  is closed because  $S_+^n$  is closed.

3.2

•  $A \in S_+^n$ ,  $u \in \mathbb{R}^n$

$u^T A u = 0$ . As  $A \in S_+^n$ ,  $\exists L \in \mathbb{R}^{n \times n}$ :  $A = L L^T$  (lower triangular)

$\Rightarrow u^T L L^T u = 0 \Rightarrow \|L^T u\|_2 = 0 \Rightarrow L^T u = 0$  by non-degeneracy of Euclidean inner product

$\therefore A u = L L^T u = L 0 = 0$

$u^T A u = 0 \Leftrightarrow u \in \ker(A)$  (other direction obvious)

•  $A \in S^n$ ,  $R$  invertible  $n \times n$

$A \succeq 0$ :  $x^T R^T A R x = (x^T R^T) A (R x) = (R x)^T A (R x) \geq 0 \quad \forall x \in \mathbb{R}^n$

$\Rightarrow R^T A R \succeq 0$

$R^T A R \succeq 0$ :  $x^T A x = x^T (R R^{-1})^T A (R R^{-1}) x = (R^{-1} x)^T R^T A R (R^{-1} x) \geq 0 \quad \forall x \in \mathbb{R}^n$

$\Rightarrow A \succeq 0$

$A \succ 0 \Leftrightarrow R^T A R \succ 0$  proceeds similarly

•  $\begin{bmatrix} A & B \\ B^T & C \end{bmatrix} \succ 0 \Rightarrow \begin{pmatrix} x^T & y^T \end{pmatrix} \begin{bmatrix} A & B \\ B^T & C \end{bmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = x^T A x + x^T B y + y^T B^T x + y^T C y > 0 \quad \forall x \in \mathbb{R}^n, y \in \mathbb{R}^m$

$A \in \mathbb{R}^{n \times n}$ ,  $C \in \mathbb{R}^{m \times m}$

$B \in \mathbb{R}^{n \times m}$

By setting  $y = 0$ ,

$\Rightarrow x^T A x > 0 \quad \forall x \in \mathbb{R}^n \Rightarrow A \succ 0$

Now set  $x = -A^{-1} B y$ , ( $A$  is invertible as all eigenvalues positive)

$y^T B^T (A^{-1})^T A A^{-1} B y \leq y^T B^T (A^{-1})^T B y \leq y^T B^T A^{-1} B y + y^T C y > 0$

$\Rightarrow y^T C y - y^T B^T A^{-1} B y > 0 \quad \forall y \in \mathbb{R}^m$

$\Rightarrow C - B^T A^{-1} B \succ 0$

For the inverse consider  $\begin{pmatrix} x^T & y^T \end{pmatrix} \begin{bmatrix} A & B \\ B^T & C \end{bmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = x^T A x + x^T B y + y^T B^T x + y^T C y \quad \forall x \in \mathbb{R}^n, y \in \mathbb{R}^m$

Now  $C - B^T A^{-1} B \succ 0$  implies as above.

Write  $x = x' - A^{-1} B y$ ,

r.h.s:  $(x' - A^{-1} B y)^T A (x' - A^{-1} B y) + x'^T B y + y^T B^T x' - y^T B^T (A^{-1})^T B y$   
 $- y^T B^T A^{-1} B y + y^T C y$

$= (x'^T A x' - y^T B^T (A^{-1})^T A x' - x'^T A A^{-1} B y + x'^T B y + y^T B^T x' \quad = S_1)$

$+ (y^T B^T (A^{-1})^T A A^{-1} B y - y^T B^T (A^{-1})^T B y - y^T B^T A^{-1} B y + y^T C y \quad = S_2)$

$S_2 > 0$  as  $C - B^T A^{-1} B > 0$  (following steps above in inverse)  $\forall y \in \mathbb{R}^m$

$S_1 = x'^T A x' > 0$  as  $A > 0$  (as  $(A^{-1})^T = A^{-1}$ )  $\forall x' \in \mathbb{R}^n \Rightarrow \forall x \in \mathbb{R}^n$

$$\therefore \begin{bmatrix} A & B \\ B^T & C \end{bmatrix} > 0$$

•  $A > B > 0$  then  $A^{-1} < B^{-1}$ .

Proof:  
Start  $B = I$ ,

$$A > B = I \Rightarrow A - I > 0 \Rightarrow x^T A x > x^T x \quad \forall x \in \mathbb{R}^n$$

Define  $x = A^{-1/2} y$ , ( $A^{-1/2}$  is well defined as  $A > 0 \Rightarrow A^{-1} > 0$ )

$$\Leftrightarrow y^T A^{-1/2} A A^{-1/2} y = y^T y > y^T A^{-1} y \quad \forall y \in \mathbb{R}^n \quad (\text{as } \ker(A) = \emptyset)$$

} commute as simultaneously diagonalizable

$$\therefore A^{-1} < I$$

Now  $A > B \Leftrightarrow B^{-1/2} A B^{-1/2} > I$  by a similar argument. Thus,

$$\Leftrightarrow (B^{-1/2} A B^{-1/2})^{-1} < I \Leftrightarrow (B^{-1})^{-1/2} A^{-1} (B^{-1})^{-1/2} < I$$

and therefore applying a similar argument yet again,

$$\Leftrightarrow A^{-1} < B^{-1}$$

•  $A, B \in S^n$ ,  $A \geq 0$ ,  $B \geq 0 \Rightarrow A \otimes B \geq 0$

Proof:

$A, B$  have the eigenvalue decompositions ( $i, j = 1, \dots, n$ ,  $k, m = 1, \dots, n$ )

$$A_{ij} = \sum_k \lambda^{(k)} v_i^{(k)} v_j^{(k)} \quad \text{where } \lambda^{(k)} \geq 0 \quad \forall k \quad A \geq 0$$

$$B_{ij} = \sum_m \mu^{(m)} u_i^{(m)} u_j^{(m)} \quad \text{where } \mu^{(m)} \geq 0 \quad \forall m \quad \text{as } B \geq 0$$

$$\Rightarrow (A \otimes B)_{ij} = A_{ij} B_{ij} = \sum_{k,m} \lambda^{(k)} \mu^{(m)} v_i^{(k)} u_i^{(m)} v_j^{(k)} u_j^{(m)}$$

Identify  $w^{(km)} := v^{(k)} \otimes u^{(m)}$  as a (not unit-normed) vector with  $w_i^{(km)} = v_i^{(k)} u_i^{(m)}$ .

$$w^{(km)} \in \mathbb{R}^n$$

Now

$$\begin{aligned} x^T (A \otimes B) x &= \sum_{i,j} x_i (A \otimes B)_{ij} x_j = \sum_{i,j,k,m} \lambda^{(k)} \mu^{(m)} w_i^{(km)} x_i w_j^{(km)} x_j \\ &= \sum_{k,m} \lambda^{(k)} \mu^{(m)} \underbrace{(x^T w^{(km)})^2}_{\geq 0} \geq 0 \quad \forall x \in \mathbb{R}^n \quad \therefore (A \otimes B) \geq 0 \end{aligned}$$