

Lec. 17 $\Sigma(\text{Ellipsoids } (\Sigma))$ 

$$\Sigma(A, \underline{b}, c) := \left\{ \underline{x} \in \mathbb{R}^n \mid \underbrace{\underline{x}^\top A \underline{x} + 2\underline{x}^\top \underline{b} + c \leq d}_{\text{convex quadratic}} \right\}$$

$$A \in \mathbb{S}^n_{++}$$

$$\underline{b} \in \mathbb{R}^n$$

$$c \in \mathbb{R}$$

$$\boxed{\underline{b}^\top A^{-1} \underline{b} - c > 0}$$

$\Sigma$  is non-empty

$$\text{Vol}(\Sigma(A, \underline{b}, c)) = \text{const}(n) \cdot \sqrt{\det((\underline{b}^\top A^{-1} \underline{b} - c) A^{-1})}$$

$$(\text{Vol}(\cdot))^2 \propto \det(A^{-1})$$

$(A, \underline{b}, c)$  parameterization of  $\Sigma$

# $(\underline{a}, Q)$ parameterization of Ellipsoid $\Sigma$

$$\Sigma(\underline{a}, Q) = \left\{ \underline{x} \in \mathbb{R}^n \mid (\underline{x} - \underline{a})^T Q^{-1} (\underline{x} - \underline{a}) \leq 1 \right\}$$

$\begin{matrix} \underline{a} \\ \mathbb{R}^n \end{matrix}$   $\begin{matrix} Q \\ S^n_{++} \end{matrix}$   $\begin{matrix} \underline{a} = \text{center vector} \\ Q = \text{shape matrix} \end{matrix}$

Sq. roots of  $\text{eig}(Q) = \text{lengths of semi-axes}$   
of  $\Sigma$

$$\text{Vol}(\Sigma(\underline{a}, Q)) = \frac{\text{Vol}(B_1^n)}{\sqrt{\det(Q^{-1})}} = \frac{\pi^{n/2}}{\Gamma(n/2 + 1)} \sqrt{\det(Q)}$$

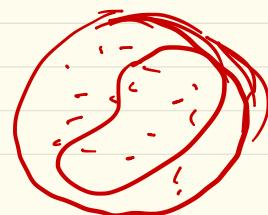
center position

$\Leftrightarrow \text{Vol}(\Sigma(\underline{a}, Q)) \propto \sqrt{\det(Q)}$



$$\frac{\text{Vol}(\text{cube})}{\text{Vol}(B_1^n)}$$

shape matrix



$$\frac{\alpha^2}{\alpha} = 1 \Rightarrow \alpha = \sqrt{2}$$

$\alpha_1, \alpha_2, \dots, \alpha_n$

$$\text{Success prob.} = \frac{\text{Vol(Cube)}}{\text{Vol}(B_1^n)} = \frac{(\sqrt{2})^n}{\pi^{n/2}} \Gamma\left(\frac{n}{2} + 1\right)$$

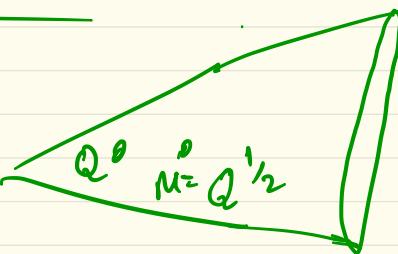
Yet another parameterization of Ellipsoid

$$S(a, M) = \left\{ M \underline{v} + \frac{a}{\sqrt{2}} \quad | \underline{v} \in \mathbb{R}^n, \|\underline{v}\|_2 \leq 1 \right\}$$

Affine image of unit ball      rotation and scaling      translation

Relation between  $(M \& Q)$ :

$$M = Q^{\frac{1}{2}}$$



Relation between  $(A, \underline{b}, c) \leftrightarrow (\underline{a}, Q)$

$$A = Q^{-1}, \quad \underline{b} \in \mathbb{R}^n$$

$$\Sigma(A, \underline{b}, c) \leftrightarrow \Sigma(\underline{a}, Q) \leftrightarrow \Sigma(\underline{a}, M)$$

$$\leftrightarrow \Sigma(\underline{a}, P) = \{\underline{x} \in \mathbb{R}^n \mid P(\underline{x} - \underline{a}) \leq \beta\}$$

Suppose,

$$Q^{-1} = PP^T, \quad P \succ 0$$

$$\Rightarrow Q = P^{-2}$$

$$\Rightarrow \sqrt{\det(Q)} = \sqrt{(\det(P^{-1}))^2}$$

$$\Rightarrow \text{vol}(\Sigma(\underline{a}, Q)) \propto \det(P^{-1}) = +(\det(P))$$

$$\Rightarrow \log \text{vol}(\Sigma(\underline{a}, Q)) \propto -\log \det(P)$$

# Minimum Vol<sup>m</sup> Outer Ellipsoid

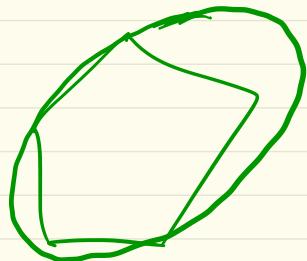
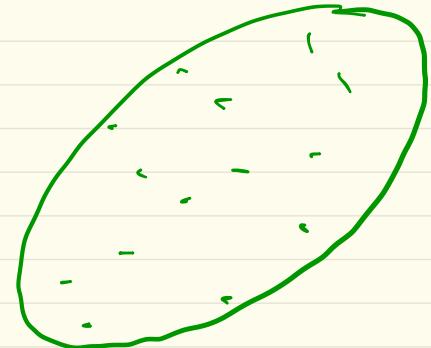
(MVOE)

also known as

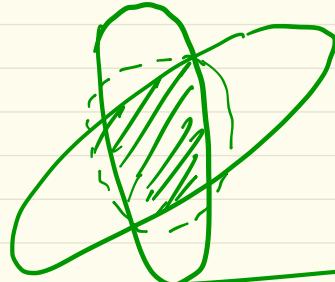
Löwner-John Ellipsoid

Fritz John  
(1948)

For any compact set  $\delta \subset \mathbb{R}^n$ ,  
the MVOE is unique



John's Thm  
This is inclusion  
tight



$$\Sigma_{LJ} = \Sigma_{MVOE}$$

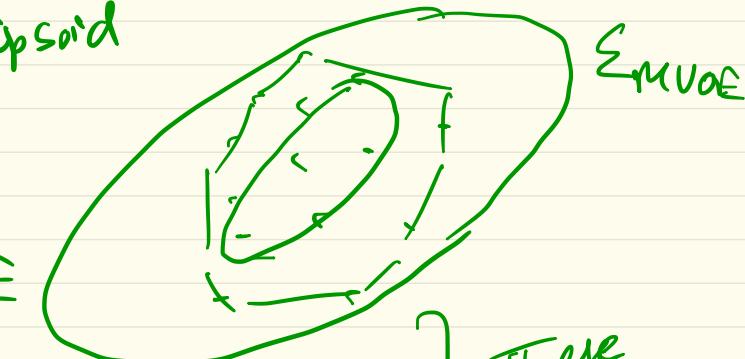
$$\boxed{\frac{1}{n} \Sigma_{MVOE} \subset \delta \subset \Sigma_{MVOE}}$$

If  $\delta$  is symmetric ( $\Leftrightarrow \underline{x} \in \delta \Rightarrow -\underline{x} \in \delta$ )  
 then the factor  $\sqrt{n}$  can be improved:

$$\frac{1}{\sqrt{n}} \sum_{MVOE} \subseteq \delta \subseteq \sum_{MVOE}$$

John's Theorem  
 This inclusion  
 is tight

Max. volume Inner Ellipsoid  
 (MVIE)



$$MVIE \subseteq \delta \subseteq n MVIE$$

for any compact  $\delta \subset \mathbb{R}^n$

If  $\delta$  is symmetric, again we have :

$$MVIE \subseteq \delta \subseteq \sqrt{n} MVIE$$

These results  
 are tight

Computing  $\mathbb{E}_{\text{MVOE}} = \mathbb{E}_{LJ}$

For arbitrary  $\delta$  (say, convex) without loss of generality

$$\min_{P > 0} \log \det(P^{-1})$$

$$P > 0$$

s.t.

$$\delta \subseteq \mathcal{E}(\underline{\alpha}, P)$$

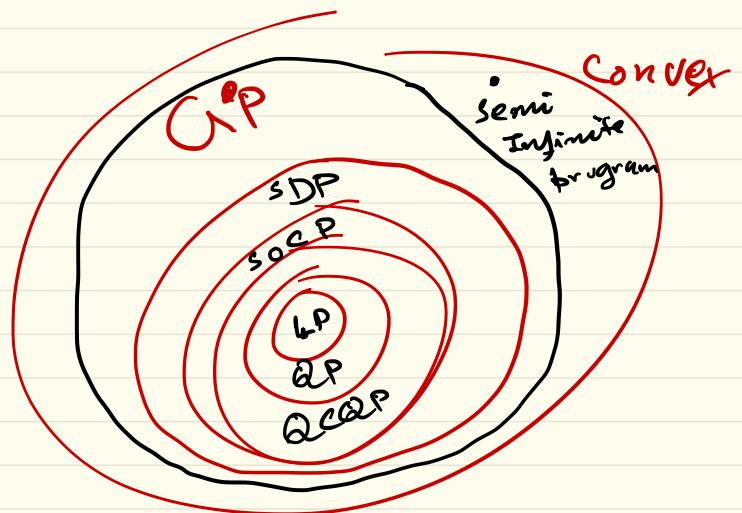
$$\sup_{x \in \delta} \|P(x - \underline{\alpha})\|_2 \leq 1$$

( $\because$  if non-convex set, then take the convex hull)

MVOE problem

Semimfinite  
programming

$LP \subseteq QP \subseteq QCQP \subseteq SOCP \subseteq SDPC \stackrel{\text{Semi infinite Programs}}{\subseteq} \text{Convex}$



Example: Suppose  $\mathcal{S}$  is finite set

$$\mathcal{S} = \left\{ \frac{\underline{a}_1}{R^n}, \dots, \frac{\underline{a}_m}{R^n} \right\}$$

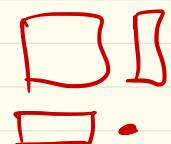
m points in  $R^n$

Our MVE

problem  
becomes :

$$\min_{P \succ 0} \log \det(P^{-1})$$

$$\text{s.t. } \left[ \| P(\underline{a}_i - \underline{q}) \|_2 \leq 1 \right]$$



$$\left[ \begin{array}{cc} I & P(\underline{a}_i - \underline{q}) \\ (P(\underline{a}_i - \underline{q}))^\top & 1 \end{array} \right] \geq 0, \quad i=1, \dots, m$$

Schur complement

$M_i(P, \underline{q})$

$$\text{diag}(M_1(P, q), \dots, M_m(P, q)) \geq 0.$$

Interior-point algorithm (can be done in cvx / cvxpy)

Computing MVE

(Max. Vol<sup>m</sup> Inner Ellipsoid)

Suppose  $\mathcal{C}$  is a convex compact set.

$$\mathcal{E}(\underline{q}, M) := \left\{ M\underline{v} + \underline{q} \mid \|\underline{v}\|_2 \leq 1 \right\},$$

$$\text{vol}(\mathcal{E}) \propto \det(M)^{-1/2} \quad M \succ 0.$$

$$\max_{M \succ 0} \text{vol}(\mathcal{E}) \propto \max_{M \succ 0} \det(M) \Leftrightarrow \max_{M \succ 0} \log \det(M)$$

Therefore, MUIE problem:

$$\max \log \det(M)$$
$$M > 0$$

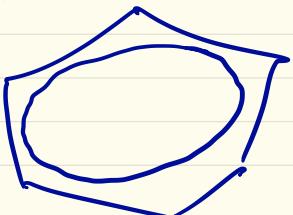
$$\text{s.t. } \underbrace{\mathcal{E}(u, M)}_{\Downarrow} \subseteq \mathcal{S}$$

$$\sup_{\|u\|_2 \leq 1} \mathbb{1}_{\mathcal{E}}(Mu + u) \leq 0$$

Example: Suppose } MUIE inside

$$\mathcal{E} = \{x \in \mathbb{R}^n \mid a_i^T x \leq b_i, i=1, \dots, m\}$$

Polyhedron



$$\boxed{\max_{M \succ 0} \log \det(M)}$$

$$\Sigma(M, \underline{a}) = \left\{ M \underline{v} + \underline{a} \mid \|\underline{v}\|_2 \leq 1 \right\}$$

$$\text{s.t. } \Sigma \subseteq \mathcal{L}$$

$$\sup_{\underline{x} \in \Sigma} \underline{a}_i^T \underline{x} \leq b_i, \quad i=1, \dots, m$$

$$\uparrow$$

$$\sup_{\|\underline{v}\|_2 \leq 1} \underline{a}_i^T (M \underline{v} + \underline{a}) \leq b_i, \quad i=1, \dots, m$$

$$\uparrow$$

$$\sup_{\|\underline{v}\|_2 \leq 1} \underline{a}_i^T M \underline{v} + \underline{a}_i^T \underline{a} \leq b_i, \quad i=1, \dots, m$$

$$\uparrow$$

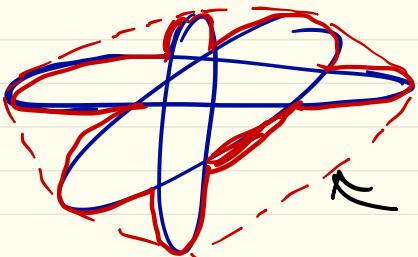
$$\boxed{\|\underline{M} \underline{a}_i\|_2 + \underline{a}_i^T \underline{a} \leq b_i, \quad i=1, \dots, m}$$

$$\begin{array}{c}
 \text{St.} \\
 \left[ \begin{array}{cc}
 (b_i - \underline{\alpha^T \alpha}) I & M \alpha_i \\
 \underline{\alpha^T M} & (b_i - \underline{\alpha_i^T \alpha}) \\
 \end{array} \right] \succ_0
 \end{array}$$

$i=1, \dots, m$

Can be done in Cvx/Cvxpy.

Union of Ellipsoids (Envvol can be computed)



Convex hull of the union

Computing EMV<sub>QE</sub> for  $\bigcup_{i=1}^m \Sigma_i$  makes use of the so-called "S-procedure".

Main idea: When is one quadratic inequality a consequence of another (or a set of other) quadratic inequalities?

S-procedure for two quadratic forms:

Suppose  $F_0, F_1 \in S^n$ .

When is it true that  $\forall z \in \mathbb{R}^n$ , the inequality

$$z^T F_1 z \geq 0 \Rightarrow z^T F_0 z \geq 0 ?$$

In words,  
when does non-negativity of one quadratic form imply non-negativity of others?

Sufficient condition:  $\exists \gamma > 0$  s.t.  $F_0 \geq \gamma F_1$

then

$$\underline{z}^T F_1 \underline{z} \geq 0 \Rightarrow \underline{z}^T F_0 \underline{z} \geq 0$$

Proof (easy):

$$F_0 \geq \gamma F_1 \Rightarrow \underline{z}^T F_0 \underline{z} \geq \underbrace{\gamma \underline{z}^T F_1 \underline{z}}_{\geq 0} \geq 0$$

$\geq 0 \text{ . (Proved)}$

Fact: The converse holds (i.e., the condition is also necessary), provided  $\exists \underline{u} \in \mathbb{R}^n$  s.t.

$$\underline{u}^T F_1 \underline{u} > 0$$

Proof: Not easy

S-procedure for two quadratic forms with strict inequalities

When does  $\underline{z}^T F_1 \underline{z} \geq 0$ ,  $\underline{z} \neq 0 \Rightarrow \underline{z}^T F_0 \underline{z} > 0$

Answer: Again,

if  $\exists \gamma \geq 0$  s.t.  $F_0 > \gamma F_1$   
then the implication holds.

Fact: Again, converse holds provided  $\exists \underline{u} \in \mathbb{R}^n$  s.t.  
 $\underline{u}^T F_1 \underline{u} > 0$ .

Slightly general version in Text, Appendix B.2 for  
quadratic forms  $x^T A_i x + 2 b_i^T x + c_i$ , etc.  
 $i=1, 2$

Also see Example B.1 there (p. 655 in text)

Back to computing  $\Sigma_{\text{MVOE}} \supseteq \bigcup_{i=1}^m \Sigma_i$ .

Suppose

$$\Sigma_i = \left\{ \underline{x} \in \mathbb{R}^n \mid \underline{x}^T A_i \underline{x} + 2 \underline{b}_i^T \underline{x} + c_i \leq 0 \right\}, \quad i=1, \dots, m$$

where  $A_i \in \mathbb{S}_{++}^n$  &  $i=1, \dots, m$ .

For  $\Sigma_{\text{MVOE}}$ , we do the  $(\underline{q}, P)$  parameterization.

Recall:  $\Sigma_{\text{MVOE}}(\underline{q}, P) = \left\{ \underline{x} \in \mathbb{R}^n \mid \|P(\underline{x} - \underline{q})\|_2 \leq 1 \right\}$

$$\Leftrightarrow \Sigma_{\text{MVOE}}(\underline{r}, P) = \left\{ \underline{x} \in \mathbb{R}^n \mid \|P\underline{x} + \underline{r}\|_2 \leq 1 \right\}$$

where  $\underline{r} := -P\underline{q} \in \mathbb{R}^n$

So we have  $\Sigma_i = \Sigma(A_i, \underline{b}_i, c_i)$ ,  $i=1, \dots, m$   
and  $\Sigma_{\text{MVOE}} = \Sigma(\underline{r}, P)$

By "S-procedure",

$$\varepsilon_i \in \Sigma_{\text{MVOE}} \iff \exists \gamma \geq 0 \text{ s.t. } \begin{bmatrix} (\underline{P}^2 - \gamma A_i) & (\underline{P}\underline{r} - \gamma b_i) \\ (\underline{P}\underline{r} - \gamma b_i)^T & \underline{r}^T \underline{r} - 1 - \gamma c_i \end{bmatrix} \leq 0.$$

(if and only if)

Since  $\text{vol}(\Sigma_{\text{MVOE}}) \propto \det(\underline{P}^{-1})$ , therefore,

minimize  $\log \det(\underline{P}^{-1})$

s.t.  $\gamma_1 \geq 0, \dots, \gamma_m \geq 0$

$$\begin{bmatrix} (\underline{P}^2 - \gamma_i A_i) & (\underline{P}\underline{r} - \gamma_i b_i) \\ (\underline{P}\underline{r} - \gamma_i b_i)^T & \underline{r}^T \underline{r} - 1 - \gamma_i c_i \end{bmatrix} \leq 0, \quad i=1, \dots, m.$$

Introducing new variables  $\tilde{r} := \underline{r}\gamma$ ,  $\tilde{P} := P^2$ ,  
 minimize  $\log \det(A^{-1})$

s.t.  $\gamma_1 \geq 0, \dots, \gamma_m \geq 0$

$$\left( \begin{array}{cc|c} (\tilde{P} - \gamma_i A_i) & (\tilde{r} - \gamma_i b_i) & 0 \\ (\tilde{r} - \gamma_i b_i) & (-1 - \gamma_i c_i) & \tilde{r}^T \\ \hline 0 & \tilde{P} & -\tilde{P} \end{array} \right) \leq 0, \quad i=1, \dots, m$$

$m$   
LMI's  
 $\tilde{P}, \tilde{r}, \gamma_i$

which is a convex problem in variables:

$$\tilde{P}, \tilde{r}, \gamma_1, \dots, \gamma_m$$

Affine invariance of  $\Sigma_{LJ}$  (both  $\Sigma_{MVOE}$  and  $\Sigma_{MVIE}$ )

Given compact  $\delta \subset \mathbb{R}^n$ .

Consider  $T \in \mathbb{R}^{n \times n}$  nonsingular. Then

$$\Sigma_{LJ}(T\delta) = T \Sigma_{LJ}(\delta)$$