

## International Online Seminar on Interval Methods in Control Engineering Encompassing computation of the ellipsoidal image, in the singular case

Morgan Louédec May 12<sup>th</sup> 2023

#### SENSTRA 44 STCC Context - Set propagation

#### **Principe:**

Set of initial states  $S_0 \subset \mathbb{R}^n$ Nonlinear mapping  $\boldsymbol{g} : \mathbb{R}^n \to \mathbb{R}^n$ Enclosing Set  $S_{\text{out}} \subset \mathbb{R}^n$  such that  $\boldsymbol{g}(S_0) \subseteq S_{\text{out}}$ 

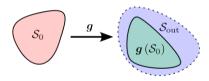


Figure 1: Set representation

#### Interest:

- guaranteed prediction algorithm
- a tool for mathematical proofs

Limitations:

- wrapping effect
- computational complexity

## Common shapes of sets

- boxes
- zonotopes
- ellipsoids

#### STICC Context - Ellipsoids

#### Definition

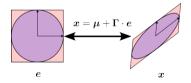
A non-degenerated ellipsoid is a subset of  $\mathbb{R}^n$  described by the quadratic form

$$\mathcal{E}(\boldsymbol{\mu},\boldsymbol{\Gamma}) = \left\{ \boldsymbol{x} \in \mathbb{R}^{n} | \left( \boldsymbol{x} - \boldsymbol{\mu} \right)^{T} \boldsymbol{\Gamma}^{-T} \boldsymbol{\Gamma}^{-1} \left( \boldsymbol{x} - \boldsymbol{\mu} \right) \leq 1 \right\}$$
(1)

with  $\Gamma \in \mathbb{R}^{n \times n}$ , the center  $\mu \in \mathbb{R}^n$  and the positive definite matrix  $\Gamma \Gamma^{T}$ .

An ellipsoid is an affine transformation of the unit sphere:

$$\mathcal{E}(\boldsymbol{\mu}, \boldsymbol{\Gamma}) = \{ \boldsymbol{x} \in \mathbb{R}^{n} | \exists \boldsymbol{e} \in \mathbb{R}^{n}, \boldsymbol{x} = \boldsymbol{\mu} + \boldsymbol{\Gamma} \cdot \boldsymbol{e}, \|\boldsymbol{e}\|_{2} \leq 1 \}$$
$$= \boldsymbol{\mu} + \boldsymbol{\Gamma} \cdot \mathcal{E}(0, \boldsymbol{I}_{n})$$
(2)



## STAR STARE S

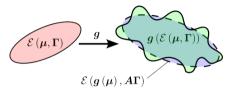


Figure 3: Estimation by linearisation

Propagation by linearisation at the point  $\mu$ :

$$\boldsymbol{\mu}_{l}=\boldsymbol{g}\left(\boldsymbol{\mu}\right) \tag{3}$$

$$\boldsymbol{\Gamma}_{l} = \boldsymbol{A} \cdot \boldsymbol{\Gamma} \tag{4}$$

$$\boldsymbol{A} = \frac{\partial \boldsymbol{g}}{\partial \boldsymbol{x}} \left( \boldsymbol{\mu} \right) \tag{5}$$

Propagation by extended Kalman filter

$$\hat{\boldsymbol{x}}_{k+1} = \boldsymbol{g}\left(\hat{\boldsymbol{x}}_{k}\right) \tag{6}$$

$$oldsymbol{G}_{k+1} = oldsymbol{A}_k \cdot oldsymbol{G}_k \cdot oldsymbol{A}_k^T + oldsymbol{G}_lpha$$
 (7)

$$\boldsymbol{A}_{k} = \frac{\partial \boldsymbol{g}}{\partial \boldsymbol{x}} \left( \hat{\boldsymbol{x}}_{k} \right) \tag{8}$$

#### Resemblance

$$\boldsymbol{\mu} \leftrightarrow \hat{\boldsymbol{x}}_k$$
 (9)

$$\boldsymbol{\Gamma}\boldsymbol{\Gamma}^{T}\leftrightarrow\boldsymbol{G}_{k}$$
 (10)

## SENSTA STICE Existing method - Theorem

#### Theorem - Rauh et al. 2022 [1]

Consider the ellipsoid  $\mathcal{E}(\mu, \Gamma)$  and the nonlinear mapping  $\boldsymbol{g}$ . The matrix  $\boldsymbol{A} = \frac{\partial \boldsymbol{g}}{\partial \boldsymbol{x}}(\mu)$  is supposed invertible. A set enclosing  $\boldsymbol{g}(\mathcal{E}(\mu, \Gamma))$  and written  $\mathcal{E}(\mu_{\text{out}}, \Gamma_{\text{out}})$  is given by

$$\boldsymbol{\Gamma}_{\text{out}} = (1+\rho) \cdot \boldsymbol{A} \cdot \boldsymbol{\Gamma} \tag{11}$$

$$oldsymbol{\mu}_{ ext{out}} = oldsymbol{g}\left(oldsymbol{\mu}
ight)$$
 (12)

where

$$\rho = \max_{\|\tilde{\boldsymbol{x}}\| \le 1} \left\| \tilde{\boldsymbol{b}}(\tilde{\boldsymbol{x}}) \right\|$$
(13)

$$\tilde{\boldsymbol{b}}(\tilde{\boldsymbol{x}}) = \boldsymbol{\Gamma}^{-1} \cdot \boldsymbol{A}^{-1} \left( \boldsymbol{g} \left( \boldsymbol{\Gamma} \cdot \tilde{\boldsymbol{x}} + \boldsymbol{\mu} \right) - \boldsymbol{g} \left( \boldsymbol{\mu} \right) \right) - \tilde{\boldsymbol{x}}$$
(14)

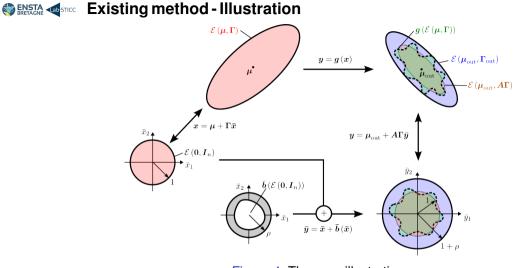
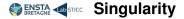


Figure 4: Theorem illustration



#### When $\boldsymbol{A}\boldsymbol{\Gamma}$ is not invertible:

- The ellipsoid  $\mathcal{E}\left(\mu_{\mathrm{out}}, \pmb{A} \pmb{\Gamma} 
  ight)$  is degenerated
- Inflating the ellipsoid won't guaranty the enclosure  $\boldsymbol{g}\left(\mathcal{E}\left(\boldsymbol{\mu},\boldsymbol{\Gamma}\right)\right)$

Common Examples:

- Projections
- Fluid damping (quadratic)

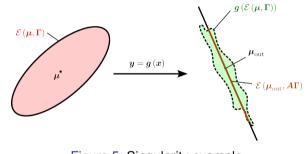


Figure 5: Singularity example

## Singularity - Degenerated ellipsoid

No quadratic form,  $\Gamma$  not invertible but there is an affine transformation:

$$\mathcal{E}\left(\mu, \Gamma
ight) = \mu + \Gamma \cdot \mathcal{E}\left(0, I_{n}
ight)$$
(15)

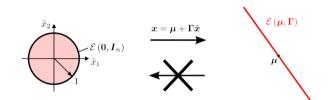


Figure 6: Degenerated example

## Singularity - Investigated solution

Start from  $\mathcal{E}(\mu_{out}, A\Gamma)$  to find an ellipsoid  $\mathcal{E}(\mu_{out}, \Gamma_s)$  similar to  $g(\mathcal{E}(\mu, \Gamma))$ 

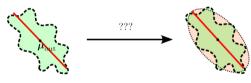


Figure 7: Find lost dimensions

Particular case, planar projection: one can keep the degenerated ellipsoid  $\mathcal{E}(\mu_{out}, \mathbf{A}\Gamma)$ 

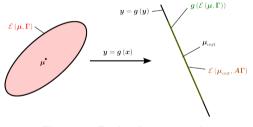


Figure 8: Projection example

### Singularity - Singular Value Decomposition

Singular value decomposition

$$\boldsymbol{A}\boldsymbol{\Gamma} = \boldsymbol{U}\cdot\boldsymbol{\Sigma}\cdot\boldsymbol{V}^{T} \qquad (16)$$

with  $\boldsymbol{U} \in \mathbb{R}^{n \times n}$  and  $\boldsymbol{V} \in \mathbb{R}^{n \times n}$  orthonormals and  $\boldsymbol{\Sigma} \in \mathbb{R}^{n \times n}$  diagonal.

The diagonal elements of  $\Sigma$  are the singular values  $\sigma_i \ge 0$ .

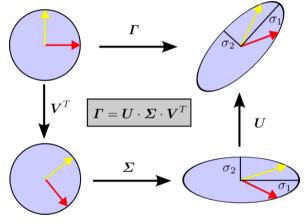


Figure 9: Example in 2 dimensions

# Singularity - Adding singular values $U^T$ Addition of singular values $oldsymbol{\Sigma} ightarrow oldsymbol{S}$ I I $\sigma_{\rm new}$ $-\mathcal{E}\left(\boldsymbol{\mu}_{ ext{out}}, \boldsymbol{US} ight)$ $-\mathcal{E}(\boldsymbol{\mu}_{\mathrm{out}}, \boldsymbol{A}\boldsymbol{\Gamma})$

Figure 10: Meet the shape of the set

## Singularity - Proposed solution

$$\mathcal{E}\left(oldsymbol{\mu}_{ ext{out}},oldsymbol{arLambda}
ight)$$
 is given by

$$\boldsymbol{\Gamma}_{\text{out}} = (1+\rho) \cdot \boldsymbol{\Gamma}_{\boldsymbol{s}} \qquad (17)$$

$$oldsymbol{\mu}_{ ext{out}} = oldsymbol{g}\left(oldsymbol{\mu}
ight)$$
 (18)

#### where

$$\rho = \max_{\|\tilde{\boldsymbol{x}}\| \le 1} \|\tilde{\boldsymbol{b}}(\tilde{\boldsymbol{x}})\|$$
(19)  
$$\tilde{\boldsymbol{b}}(\tilde{\boldsymbol{x}}) = \boldsymbol{W}(\boldsymbol{g}(\boldsymbol{\Gamma} \cdot \boldsymbol{e} + \mu) - \boldsymbol{g}(\mu)) - \boldsymbol{Z}\tilde{\boldsymbol{x}}$$
(20)

and where the matrices  $\Gamma_s$ , **W** and **Z** from  $\mathbb{R}^{n \times n}$  are given by:

• (General case), if **A***I*<sup>•</sup> is invertible, then

$$\Gamma_s = A\Gamma$$
 (21)

$$\boldsymbol{W} = \boldsymbol{\Gamma}_s^{-1} \tag{22}$$

$$\boldsymbol{Z} = \boldsymbol{I}_n \tag{23}$$

Singularity - Proposed solution

(Singular case), if *AΓ* is not invertible, then

 $\Gamma_s = US$  $W = S^*U^T$  $Z = S^*SV^T$ 

$$(24) \qquad \mathbf{s}_{i} = \begin{cases} \sigma_{i} & \text{if } \sigma_{i} > 0\\ \max_{\mathbf{x} \in \mathcal{E}(\boldsymbol{\mu}, \boldsymbol{\Gamma})} \left| \mathbf{e}_{i}^{T} \mathbf{U}^{T} \left( \mathbf{g} \left( \mathbf{x} \right) - \mathbf{g} \left( \boldsymbol{\mu} \right) \right) \right| & \text{if } \sigma_{i} = 0\\ (25) \\ (26) \qquad \mathbf{s}_{i}^{*} = \begin{cases} 1/s_{i} & \text{if } \mathbf{s}_{i} \neq 0\\ 0 & \text{else} \end{cases}$$
(29)

where

with the diagonal matrices

$$m{S} = ext{diag}(s_1, s_2, \dots, s_n)$$
 (27)  
 $m{S}^* = ext{diag}(s_1^*, s_2^*, \dots, s_n^*)$  (28)

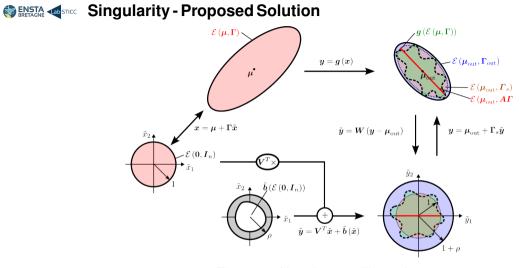


Figure 11: New theorem illustration

## SERSTA STICE Application - Consensus

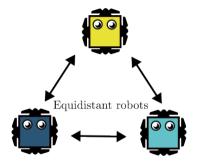


Figure 12: Example of formation control by consensus

Continuous system with three robots and periodic position measurement:

$$\dot{x}_i = v_i,$$
 (30)

$$\dot{v}_i = u_i, \tag{31}$$

$$y_{i,k} = x_i(t_k),$$
 (32)

$$t_{k} = k \cdot \delta_{t} \tag{33}$$

with  $i \in \mathcal{I}_3$ ,  $\mathcal{I}_3 = \{1, 2, 3\}$ ,  $k \in \mathbb{N}$ ,  $x_i \in \mathbb{R}$ ,  $v_i \in \mathbb{R}$ ,  $u_i \in \mathbb{R}$  and the period between the measurements  $\delta_t > 0$ 



The robots must find a consensus

$$\lim_{t \to \infty} \|x_i(t) - x_j(t)\| = 0, \quad (34)$$
$$\lim_{t \to \infty} \|v_i(t) - v_j(t)\| = 0 \quad (35)$$

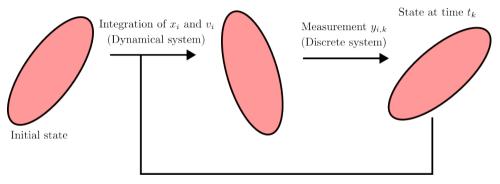
for all  $i, j \in \mathcal{I}_3$ 

#### Consensus protocol [Zheng et al.] [2]:

$$\dot{u}_{i}(t) = \sum_{j \neq i} (y_{j,k} - x_{i}(t)) - c \cdot v_{i}(t), \ t \in (t_{k}, t_{k} + 1]$$
(36)

with the feedback gain c > 0.

## SERSTA STICE Application - Propagation



#### Figure 13: Propagation process

## SENSTA STICE Application - Conditioning

#### State vector

$$\mathbf{z} = [z_i]_{i \in [1,9]}, \quad (37)$$

$$z_1(t) = x_2(t) - x_1(t), \quad (38)$$

$$z_2(t) = v_2(t) - v_1(t), \quad (39)$$

$$z_3(t) = x_3(t) - x_1(t), \quad (40)$$

$$z_4(t) = v_3(t) - v_1(t), \quad (41)$$

$$z_5(t) = x_3(t) - x_2(t), \quad (42)$$

$$z_6(t) = v_3(t) - v_2(t), \quad (43)$$

$$z_{7,k} = y_{2,k} - y_{1,k}, \quad (44)$$

$$z_{8,k} = y_{3,k} - y_{1,k}, \quad (45)$$

$$z_{9,k} = y_{3,k} - y_{2,k}. \quad (46)$$

#### Initial set of state

$$\| m{z}(0) \| < e,$$
 (47)  
 $m{z}_{6+j,0} = m{z}_j(0), \, {
m for} \, j \in [1:3]$  (48)

with e>0, equivalent to  $m{z}(0)\in \mathcal{E}\left(0, m{\Gamma}_{0}
ight)$  with the singular matrix

$$\boldsymbol{\Gamma}_{0} = \boldsymbol{e} \cdot \begin{bmatrix} \boldsymbol{I}_{6} & \boldsymbol{0}_{6,3} \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & \boldsymbol{0}_{3,3} \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$
(49)

#### STEC Application - Singuliarity

Cause of the singularity:

- The initial state is a degenerated ellipsoid
- The measurements are linear projections

By the propagation of the degenerated ellipsoid, on can find a matrix  $\Gamma_{t_{\rm end}}$  such that

 $oldsymbol{z}(t_{ ext{end}}) \in \mathcal{E}\left(0, oldsymbol{\Gamma}_{t_{ ext{end}}}
ight)$  (50)

with  $t_{end} > 0$ 

One can then verify that the system is contracting

$$t_{\mathrm{end}} > 0, \, \mathcal{E}\left(0, \boldsymbol{\Gamma}_{t_{\mathrm{end}}}\right) \subseteq \mathcal{E}\left(0, \boldsymbol{\Gamma}_{0}\right)$$
 (51)



The proposed solution in the singular case can be used to

- consider degenerated ellipsoids
- study systems with projections or other singular mappings

This tool is being tested for the stability analysis of n-dimensional non-linear hybrid systems



#### Andreas Rauh and Luc Jaulin.

A computationally inexpensive algorithm for determining outer and inner enclosures of nonlinear mappings of ellipsoidal domains.

International Journal of Applied Mathematics and Computer Science, 31(3):399–415, 2021.

Yuanshi Zheng, Qi Zhao, Jingying Ma, and Long Wang. Second-order consensus of hybrid multi-agent systems. *Systems & Control Letters*, 125:51–58, March 2019.