



Set-membership state estimation for uncertain systems based on zonotopes and ellipsoids

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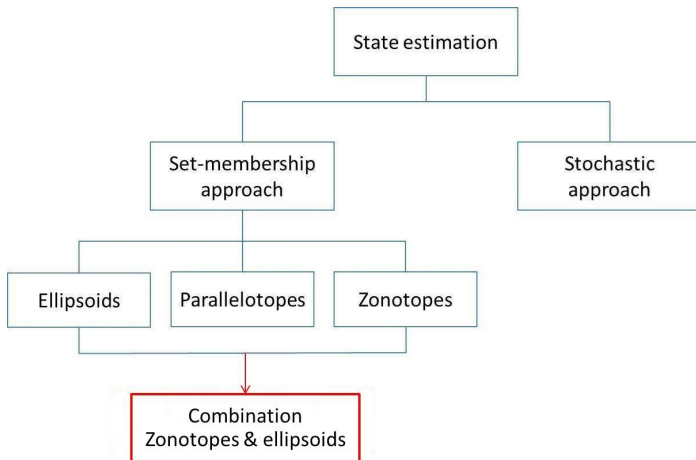
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 - Objective
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- 3 Notations and properties
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 - Correction step
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 - Prediction step
 - Correction step
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Two approaches for estimation

Knowing the state of a system is crucial for solving many control problems.



System

Consider the linear discrete-time invariant system:

$$\begin{cases} x_{k+1} = Ax_k + F\omega_k \\ y_k = c^\top x_k + \sigma v_k \end{cases} \quad (1)$$

where $x_k \in \mathbb{R}^n$, $y_k \in \mathbb{R}$, $\omega_k \in \mathbb{R}^n$ and $v_k \in \mathbb{R}$. The pair (c^\top, A) is detectable.

Objective

Assume that $x_0 \in \mathcal{X}_0$, $\omega_k \in \mathcal{W}$ and $v_k \in \mathcal{V}$ for all $k > 0$.

Set-membership approach:

$$x_k \in \hat{\mathcal{X}}_k \Rightarrow x_{k+1} \in \hat{\mathcal{X}}_{k+1}.$$

→ Find the optimal set $\hat{\mathcal{X}}_{k+1}$ that contains x_{k+1} .

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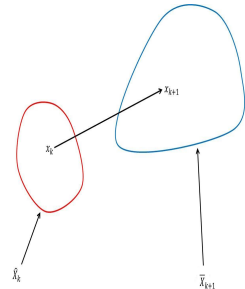
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- Prediction step

$$\bar{\mathcal{X}}_{k+1} = A\hat{\mathcal{X}}_k \cup FW$$

where $\bar{\mathcal{X}}_{k+1}$ is a predicted state estimation set.



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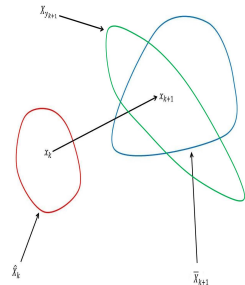
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where $\bar{\mathcal{X}}_{k+1}$ is a predicted state estimation set.

- Measurement

$$\mathcal{X}_{y_{k+1}} = \{x_{k+1} \in \mathbb{R}^n : |c^\top x_{k+1} - y_{k+1}| \leq \sigma\}$$

where $\mathcal{X}_{y_{k+1}}$ is the consistent state set by using the measurement y_{k+1} .



System

Consider the linear discrete-time invariant system:

$$\begin{cases} x_{k+1} = Ax_k + F\omega_k \\ y_k = c^T x_k + \sigma v_k \end{cases}$$

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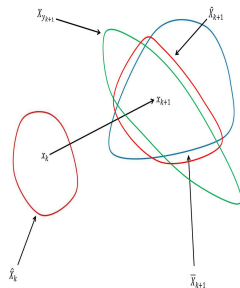
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- Correction step

$$\hat{\mathcal{X}}_{k+1} \supseteq \bar{\mathcal{X}}_{k+1} \cap \mathcal{X}_{y_{k+1}}$$

where $\hat{\mathcal{X}}_{k+1}$ is the guaranteed state estimation set.



Mathematical notations

- Interval:** $[a, b] = \{x \in \mathbb{R} : a \leq x \leq b\}$
 Unitary interval: $\mathbf{B} = [-1, 1]$
 Box (interval vector): $([a_1, b_1], \dots, [a_n, b_n])^\top \subset \mathbb{R}^n$
 Unitary box: $\mathbf{B}^n = \{x \in \underbrace{([-1, 1], \dots, [-1, 1])^\top}_{n \text{ times}}\} \subset \mathbb{R}^n$.
- Strip:** $\mathcal{S}(y, d) = \{x \in \mathbb{R}^n : |y - d^\top x| \leq 1\}$.
- Minkovsky sum:** $\mathcal{A} \oplus \mathcal{B} = \{a + b : a \in \mathcal{A}, b \in \mathcal{B}\}$.
- Ellipsoid:** $\mathcal{E}(c, P) = \{x : (x - c)^\top P^{-1}(x - c) \leq 1\}$, where $c \in \mathbb{R}^n$ is the center and a matrix $P = P^\top \succ 0$ characterizes its shape and size.
- Zonotope:** a convex symmetric polytope.
 m -zonotope: the set $\mathcal{Z} = p \oplus H\mathbf{B}^m = \{x \in \mathbb{R}^n : x = p + Hz, z \in \mathbf{B}^m\}$, with a vector $p \in \mathbb{R}^n$ and a matrix $H \in \mathbb{R}^{n \times m}$.
- P -radius of a m -zonotope $\mathcal{Z} = p \oplus H\mathbf{B}^m$:**
 $L = \max_{x \in \mathcal{Z}} \|x - p\|_P^2 = \max_{x \in \mathcal{Z}} (x - p)^\top P (x - p)$, with $P = P^\top \succ 0$

Properties

- Property 1:** Affine transformation of an ellipsoid
 $A \mathcal{E}(c, P) + b = \mathcal{E}(Ac + b, APA^\top)$, where $A \in \mathbb{R}^{n \times n}$ and $b \in \mathbb{R}^n$.
- Property 2:** Sum of two ellipsoids [Durieu:01]
 $\mathcal{E}_1(c_1, P_1) \cup \mathcal{E}_2(c_2, P_2) \subseteq \mathcal{E}(c, P)$, with $c = c_1 + c_2$, $P = \phi_1^{-1}P_1 + \phi_2^{-1}P_2$ and $\phi_1 + \phi_2 = 1$.
- Property 3:** Intersection between an ellipsoid and a strip [Fogel:82]
 $\mathcal{E}(c, P) \cap \mathcal{S}(y, d) \subseteq \mathcal{E}'(c', P')$, with $c' = c + \frac{\psi\delta}{1 + \psi g}Pd$, $\psi \geq 0$,
 $P' = (1 + \psi - \frac{\psi\delta^2}{1 + \psi g})(P - \frac{\psi}{1 + \psi g}Pdd^\top P)$, $g = d^\top Pd$ and $\delta = y - d^\top c$.
- Property 4:** Affine transformation of a zonotope [Combastel:03]
 $A\mathcal{Z} = (Ap) \oplus (AH)\mathbf{B}^m$, with $A \in \mathbb{R}^{n \times n}$.
- Property 5:** The Minkovsky sum of two centered zonotopes [Combastel:03]
 $\mathcal{Z}_1 \oplus \mathcal{Z}_2 = [H_1 \quad H_2] \mathbf{B}^{m_1+m_2}$, with $\mathcal{Z}_1 = H_1 \mathbf{B}^{m_1} \subseteq \mathbb{R}^n$ and $\mathcal{Z}_2 = H_2 \mathbf{B}^{m_2} \subseteq \mathbb{R}^n$.
- Property 6:** Intersection between a zonotope and a strip [Le:2012]
 $\mathcal{Z} \cap \mathcal{S}(\frac{d}{\sigma}, \frac{c}{\sigma}) \subseteq \hat{\mathcal{Z}}(\lambda) = \hat{p}(\lambda) \oplus \hat{H}(\lambda)\mathbf{B}^{m+1}$, with $\hat{p}(\lambda) = p + \lambda(d^\top - c^\top p)$,
 $\hat{H}(\lambda) = [(I - \lambda c^\top)H \quad \sigma\lambda]$ and $\lambda \in \mathbb{R}^n$.

System

$$\begin{cases} x_{k+1} = Ax_k + F\omega_k \\ y_k = c^\top x_k + \sigma v_k \end{cases}$$

Assume that $x_0 \in \mathcal{E}_0(c_0, P_0)$, $\mathcal{W} = \mathcal{E}(0, I_n)$, $\mathcal{V} = \mathbf{B}$ and $x_k \in \mathcal{E}_k(c_k, P_k)$.

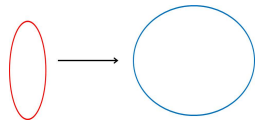
Prediction step:

$$\bar{\mathcal{E}}_{k+1}(c_{k+1}, P_{k+1}) = \hat{\mathcal{E}}_k(Ac_k, AP_k A^\top) \cup \mathcal{E}(0, FF^\top)$$

Property 2 implies that $\exists \phi = [\phi_1, \phi_2]^\top \in \mathbb{R}^2$ such that:

$$\begin{cases} c_{k+1} = Ac_k \\ P_{k+1} = \phi_1^{-1} P_1 + \phi_2^{-1} P_2 \end{cases}$$

with $P_1 = AP_k A^\top$, $P_2 = FF^\top$ and $\phi_1 + \phi_2 = 1$.



Minimization of the size of $\bar{\mathcal{E}}_{k+1}(c_{k+1}, P_{k+1})$

Trace criterion:
 [Durieu et al., 2001]

$$\phi^* = \min_{\phi} \text{tr}(P_{k+1})$$

Determinant criterion:
 [Durieu et al., 2001]

$$\phi^* = \min_{\phi} \log \det(P_{k+1})$$

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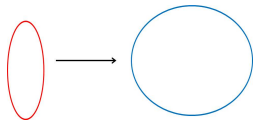
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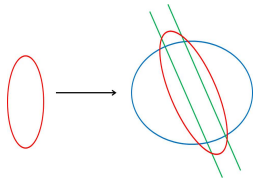
$$\hat{\mathcal{E}}_{k+1}(\hat{c}_{k+1}, \hat{P}_{k+1}) \supseteq \bar{\mathcal{E}}_{k+1}(c_{k+1}, P_{k+1}) \cap \mathcal{X}_{y_{k+1}}$$

Property 3 implies that $\exists \psi \geq 0$ such that:

$$\hat{c}_{k+1} = c_{k+1} + \frac{\psi \delta}{1 + \psi g} P_{k+1} \frac{c}{\sigma}$$

$$\hat{P}_{k+1} = (1 + \psi - \frac{\psi \delta^2}{1 + \psi g}) (P_{k+1} - \frac{\psi}{1 + \psi g} P_{k+1} \frac{c}{\sigma} \frac{c^\top}{\sigma} P_{k+1})$$

$$\text{with } g = \frac{c^\top}{\sigma} P_{k+1} \frac{c}{\sigma} \text{ and } \delta = \frac{y_{k+1}}{\sigma} - \frac{c^\top}{\sigma} c_{k+1}.$$



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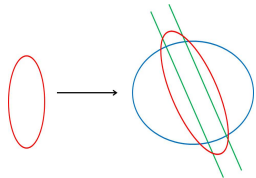
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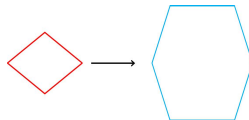
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Prediction step:

$$\bar{\mathcal{Z}}_{k+1} = A\hat{\mathcal{Z}}_k \oplus F\mathcal{W} = A\hat{p}_k \oplus \begin{bmatrix} \hat{H}_k & F \end{bmatrix} \mathbf{B}^{r+n}.$$



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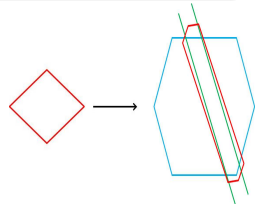
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with

$$\begin{cases} \hat{p}_{k+1}(\lambda) = A\hat{p}_k + \lambda (y_{k+1} - c^\top A\hat{p}_k) \\ \hat{H}_{k+1}(\lambda) = \begin{bmatrix} (I - \lambda c^\top) A\hat{H}_k & (I - \lambda c^\top) F \end{bmatrix} \sigma \lambda \end{cases}$$

How to compute the vector $\lambda \in \mathbb{R}^n$ in order to minimize the size of the zonotope $\hat{\mathcal{Z}}_{k+1}$?



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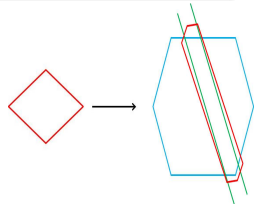
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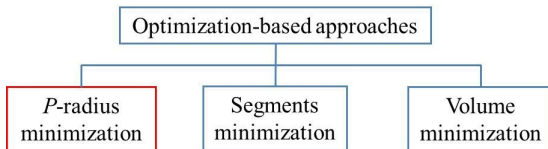
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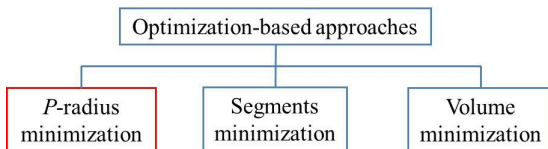
The *P*-radius of the state estimation zonotope at time k is:

$$L_k = \max_{x \in \hat{\mathcal{Z}}_k(\lambda)} \|x - \hat{p}_k\|_P^2 = \max_{x \in \hat{\mathcal{Z}}_k(\lambda)} (x - \hat{p}_k)^\top P (x - \hat{p}_k).$$

The non-increasing condition of the *P*-radius is given by [Le et al., 2013]:

$$L_{k+1} \leq \beta L_k + \max_{\omega \in \mathbf{B}^n} \|F\omega\|_2^2 + \sigma^2,$$

$$0 < \beta < 1.$$



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$$0 < \beta < 1.$$

Sufficient condition is to solve the LMI [Le et al., 2013]:
 Find the smallest value of $\beta \in (0, 1)$ such that:

$$\max_{\tau, P, Y} \tau$$

subject to

$$\left\{ \begin{array}{l} \frac{(1-\beta)P}{\sigma^2 + \text{const}} \preceq \tau I \\ \left[\begin{array}{ccc|cc} \beta P & 0 & 0 & A^\top P - A^\top c Y^\top & \\ * & F^\top F & 0 & F^\top P - F^\top c Y^\top & \\ * & * & \sigma^2 & Y^\top \sigma & \\ * & * & * & P & \end{array} \right] \preceq 0 \end{array} \right.$$

with $\text{const} = \max_{\omega \in B^n} \|F\omega\|_2^2$ and the decision variables are $\tau \in \mathbb{R}$, $P \in \mathbb{R}^{n \times n}$, and $Y = P\lambda \in \mathbb{R}^n$.

Advantages of the two previous estimation approaches:

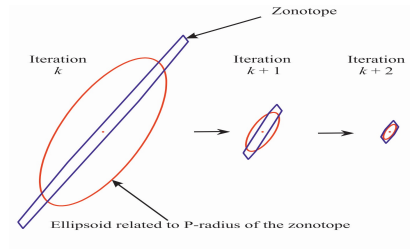
- Good estimation accuracy using the zonotopic estimation method
- Low complexity using the ellipsoidal estimation

→ The idea is to combine these two methods!

Question: How to make a transition from the zonotopic estimation to the ellipsoidal estimation?

Ellipsoid related to the P -radius of the zonotope $\hat{\mathcal{Z}}_k$:

$$\mathcal{E}(\hat{p}_k, L_k P^{-1}) = \{x_k \in \mathbb{R}^n : (x_k - \hat{p}_k)^\top (L_k P^{-1})^{-1} (x_k - \hat{p}_k) \leq 1\}.$$

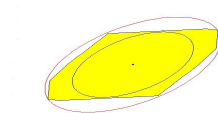


$$L_k \rightarrow L_\infty = \frac{\sigma^2 + \text{const}}{1 - \beta}$$

Scaling of the ellipsoid related to the P -radius of the zonotope $\hat{\mathcal{Z}}_k$:

$$\mathcal{E}(\hat{p}_k, L_k P^{-1}) = \{x_k \in \mathbb{R}^n : (x_k - \hat{p}_k)^\top \alpha (L_k P^{-1})^{-1} (x_k - \hat{p}_k) \leq 1\}.$$

with $\alpha \in (0, 1)$.



Objective: Obtain the smallest ellipsoid which is the outer bound of the zonotope $\hat{\mathcal{Z}}_k$!

Solution:

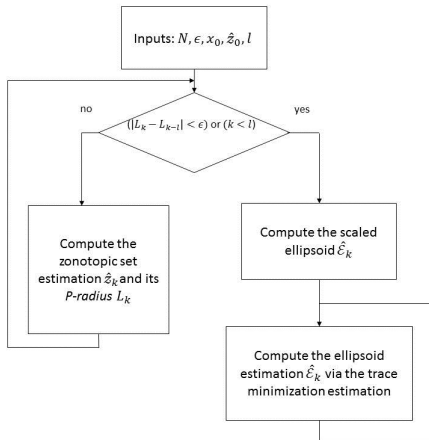
$\max \alpha$

subject to

$$\begin{cases} 0 < \alpha \leq 1 \\ (x_k - \hat{p}_k)^\top \alpha P L_k^{-1} (x_k - \hat{p}_k) \leq 1, \quad \forall x_k \in \mathcal{V}_{\hat{\mathcal{Z}}_k}, \end{cases}$$

with $\mathcal{V}_{\hat{\mathcal{Z}}_k}$ the vertices of $\hat{\mathcal{Z}}_k$.

Algorithm: Combination of the two methods [Ben Chabane et al., 2014]



where l is the length of the horizon of slow variation of the P -radius and ϵ the desired level of accuracy of the P -radius

Example

Consider the following linear discrete-time invariant system:

$$\begin{cases} x_{k+1} = \begin{bmatrix} 1 & 1 \\ 0 & 0.8 \end{bmatrix} x_k + \begin{bmatrix} -0.24 \\ 0.04 \end{bmatrix} \omega_k \\ y_k = \begin{bmatrix} -2 & 1 \end{bmatrix} x_k + 0.4 v_k \end{cases}$$

with $\|v_k\|_\infty \leq 1$, $\|\omega_k\|_\infty \leq 1$. The initial state belongs to the box $3\mathbf{B}^2$.

Example

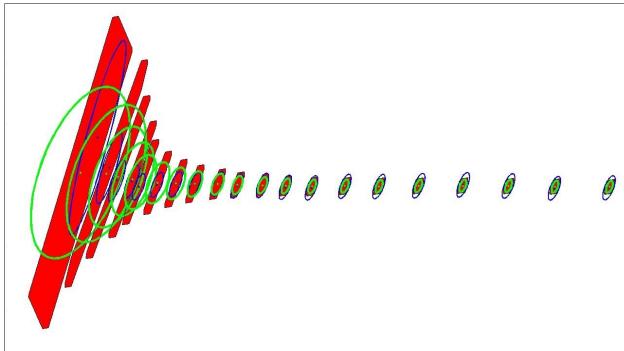


Figure: State space sets

Example

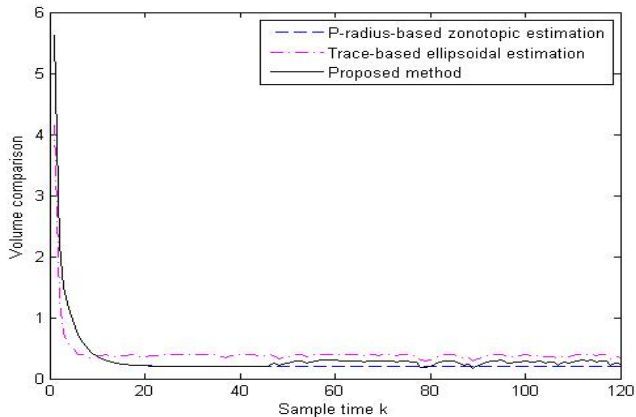


Figure: Comparison of the volumes

Example

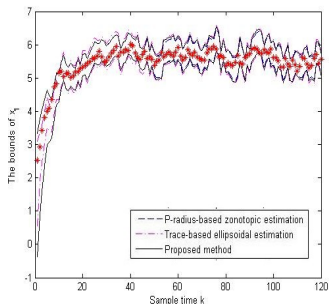


Figure: Bounds of x_1 using the three methods

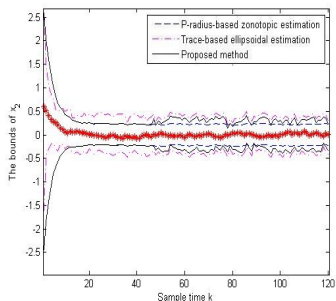


Figure: Bounds of x_2 using the three methods

Example

Table: Total computation time after 120 time instants

Algorithm	Time(second)
P -radius based zonotopic estimation	0.2652
Trace-based ellipsoidal estimation	0.0808
Combined method	0.1479

Conclusion:

- Set-membership estimation based on zonotopes is more accurate than ellipsoidal estimation but with higher complexity
- A new criterion based on the P -radius is proposed in order to make a transition from the zonotopic estimation to the ellipsoidal estimation
- The proposed method offers a good trade-off between the accuracy and the complexity

Perspectives:

- Apply the proposed method to Fault Detections (FD) and Fault Tolerant Control (FTC) purposes

**THANK YOU FOR YOUR
ATTENTION**

2014



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