

ROPO Extreme.

A method in Set-membership State Estimation

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Motivation

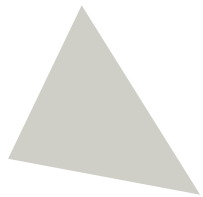
Given the following dynamic state,

$$\mathbf{x}_{k+1} = \mathbf{A}\mathbf{x}_k + \mathbf{B}\mathbf{u}_k + \mathbf{E}\boldsymbol{\omega}_k$$

Inaccuracy in Sensors

$$\mathbf{y}_k = \mathbf{C}\mathbf{x}_k + \mathbf{F}\boldsymbol{\nu}_k$$

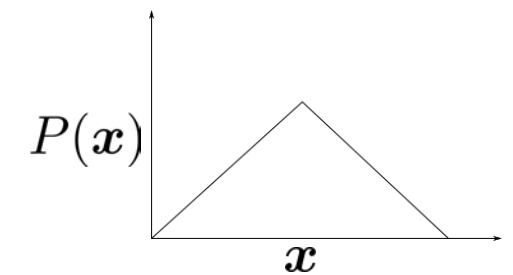
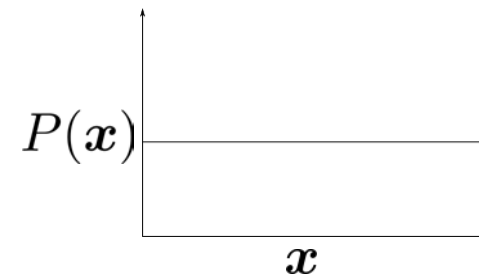
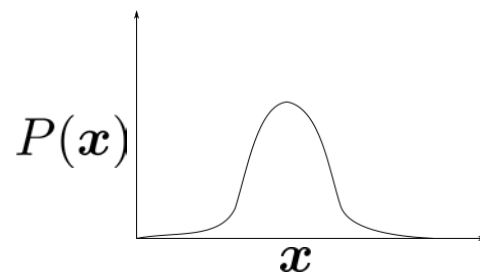
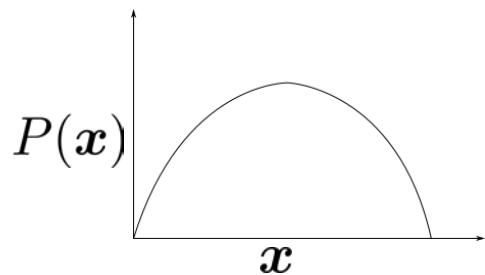
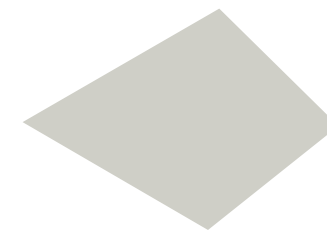
States
 $\mathbf{x}_k \in \mathcal{X}_k$



Uncertainties
and nonlinearities
 $\boldsymbol{\omega}_k \in \Omega$



Noise
 $\boldsymbol{\nu}_k \in \Psi$



There is no way to get the value of \mathbf{x}_k .

How to design an observer that guarantees the state of the system?

Motivation

The Robust Estimation Problem.

- EKF, Luenberger Observer, MHE, SSE, etc.

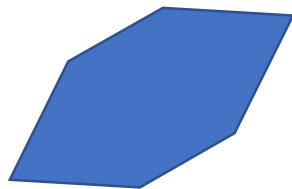
Set-membership State Estimation.

- Needs only measurement and uncertainties bounds.

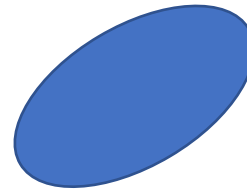
- Parallelotopes.



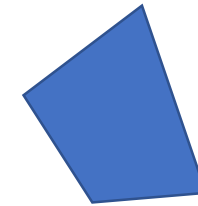
- Zonotopes.



- Ellipsoids.



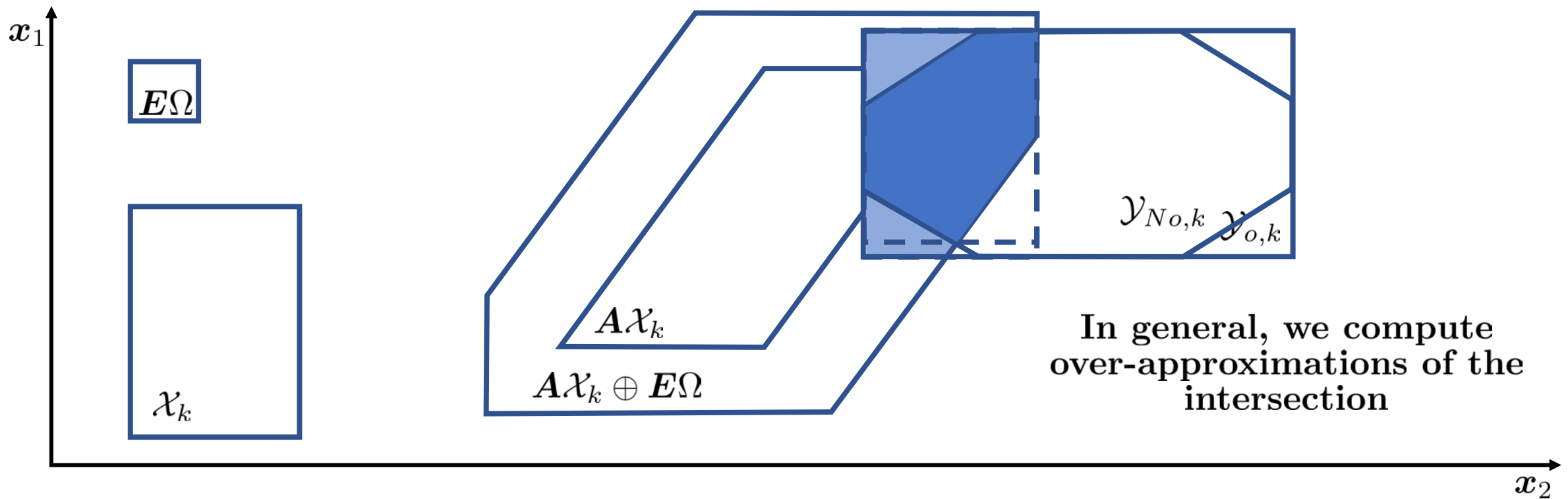
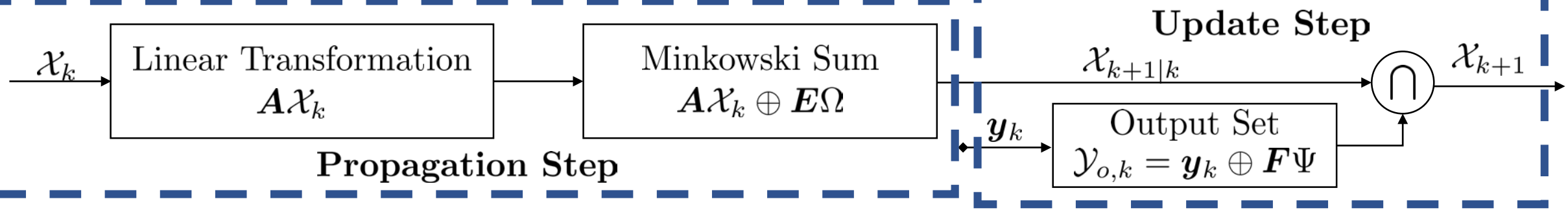
- Polytopes.



All set-based estimation techniques require a trade-off between

- Accuracy.
- Low Complexity and Computational Cost.

Set-membership State Estimation (SSE)



Parallelotopes

Generator Form

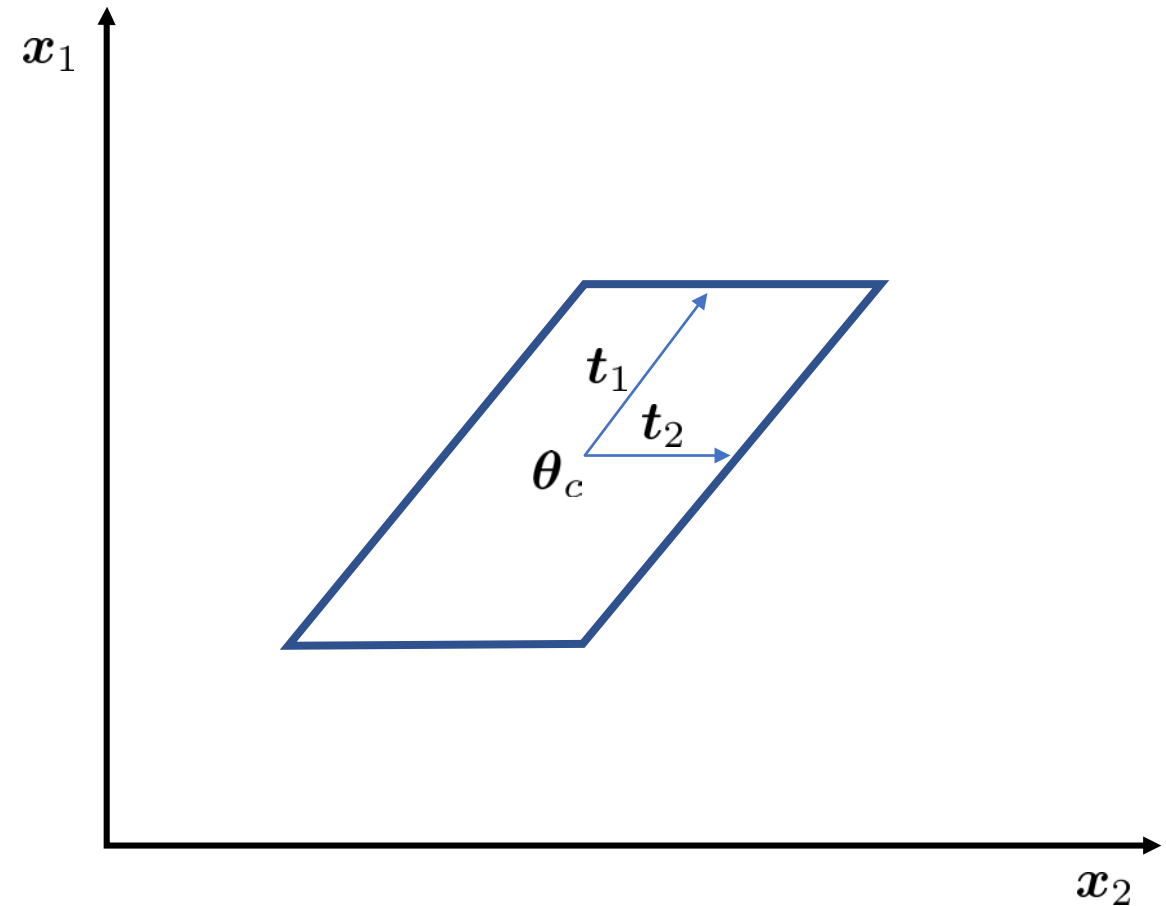
$$\mathcal{P}(\mathbf{T}, \boldsymbol{\theta}_c) = \{\mathbf{x} = \mathbf{T}\mathbf{v} + \boldsymbol{\theta}_c, \|\mathbf{v}\|_\infty \leq \mathbf{1}\}$$

\mathbf{T} – Generator matrix $[\mathbf{t}_1 \ \mathbf{t}_2 \ \dots \ \mathbf{t}_n]$

$\boldsymbol{\theta}_c$ – Center of a parallelotope.

Half-space Representation

$$\mathcal{P}(\mathbf{G}_p, \mathbf{h}_p) = \left\{ \mathbf{x} \in \mathbb{R}^n \mid \begin{bmatrix} \mathbf{P} \\ -\mathbf{P} \end{bmatrix} \mathbf{x} \leq \begin{bmatrix} \mathbf{1} + \mathbf{c} \\ \mathbf{1} - \mathbf{c} \end{bmatrix} \right\}$$

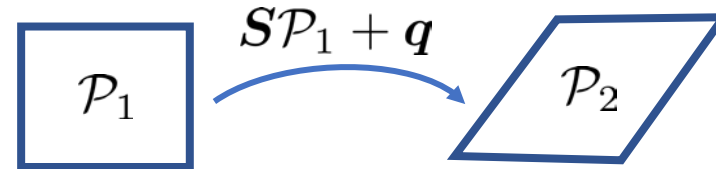


Parallelotopes

Linear Transformation

$$\mathcal{P}_2(\mathbf{T}_2, \boldsymbol{\theta}_{c,2}) = \mathbf{S}\mathcal{P}_1(\mathbf{T}_1, \boldsymbol{\theta}_{c,1}) + \mathbf{q}$$

$$\mathbf{T}_2 = \mathbf{S}\mathbf{T}_1 \quad \boldsymbol{\theta}_{c,2} = \mathbf{S}\boldsymbol{\theta}_{c,1} + \mathbf{q}$$

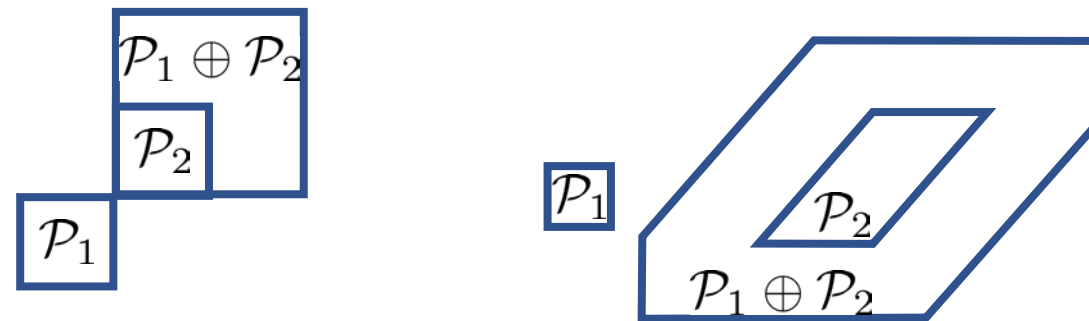


Minkowski sum

$$\mathcal{P}_1(\mathbf{T}_1, \boldsymbol{\theta}_{c,1}) \oplus \mathcal{P}_2(\mathbf{T}_2, \boldsymbol{\theta}_{c,2})$$

$$\mathbf{T}_s = [\mathbf{T}_1 \quad \mathbf{T}_2] \quad \boldsymbol{\theta}_{c,s} = \boldsymbol{\theta}_{c,1} + \boldsymbol{\theta}_{c,2}$$

This is not a parallelotope



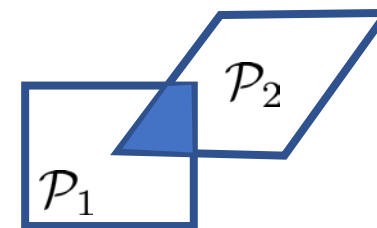
Intersection

$$\mathcal{P}_1(\mathbf{T}_1, \boldsymbol{\theta}_{c,1}) \cap \mathcal{P}_2(\mathbf{T}_2, \boldsymbol{\theta}_{c,2})$$

$$\mathbf{T}_\cap = 2\mathbf{T}_1(\mathbf{T}_1 + \mathbf{T}_2)^{-1}\mathbf{T}_2$$

$$\boldsymbol{\theta}_{c,\cap} = \mathbf{T}_2(\mathbf{T}_1 + \mathbf{T}_2)^{-1}\boldsymbol{\theta}_{c,1} + \mathbf{T}_1(\mathbf{T}_1 + \mathbf{T}_2)^{-1}\boldsymbol{\theta}_{c,2}$$

Our overapproximation proposition



Strips

$$\mathcal{S}(\mathbf{p}, c) := \{\mathbf{x} \mid \mathbf{p}^\top \mathbf{x} - c \leq 1\}$$

$$\mathcal{H}^+ = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{p}^\top \mathbf{x} = c + 1\}$$

$$\mathcal{H}^- = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{p}^\top \mathbf{x} = c - 1\}$$

Strip Propagation

$$\mathbf{p}_{k+1} := (\mathbf{p}_k^\top \mathbf{A}^{-1})^\top, \quad c_{k+1} := c_k + \mathbf{p}_k^\top \mathbf{A}^{-1} \mathbf{B} \mathbf{u}.$$

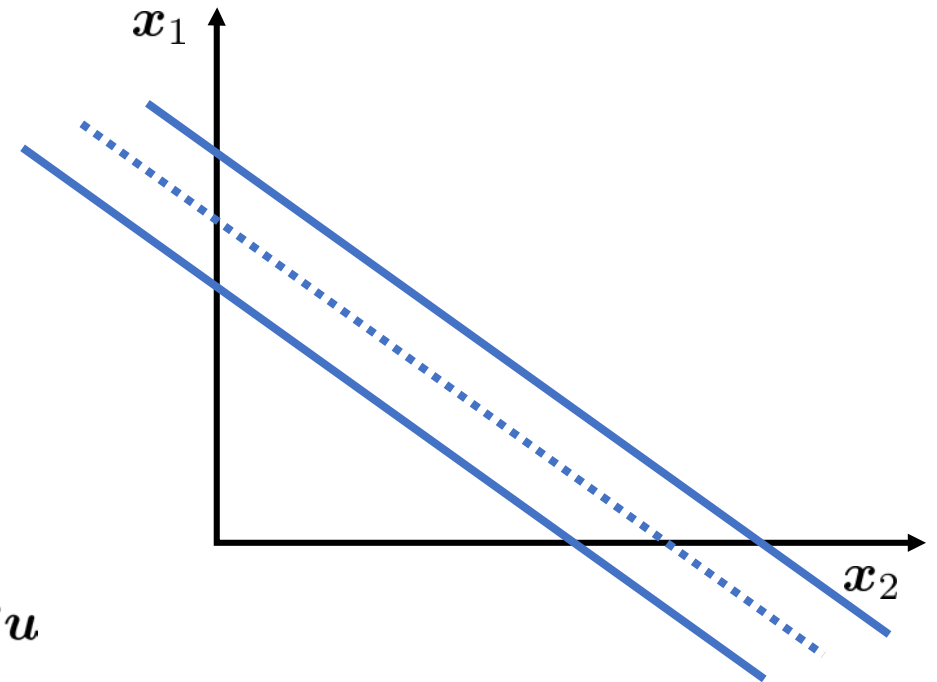
proof:

$$\mathbf{x}_{k+1} = \mathbf{A} \mathbf{x}_k + \mathbf{B} \mathbf{u} \quad \rightarrow \quad \mathbf{x}_k = \mathbf{A}^{-1} \mathbf{x}_{k+1} - \mathbf{A}^{-1} \mathbf{B} \mathbf{u}$$

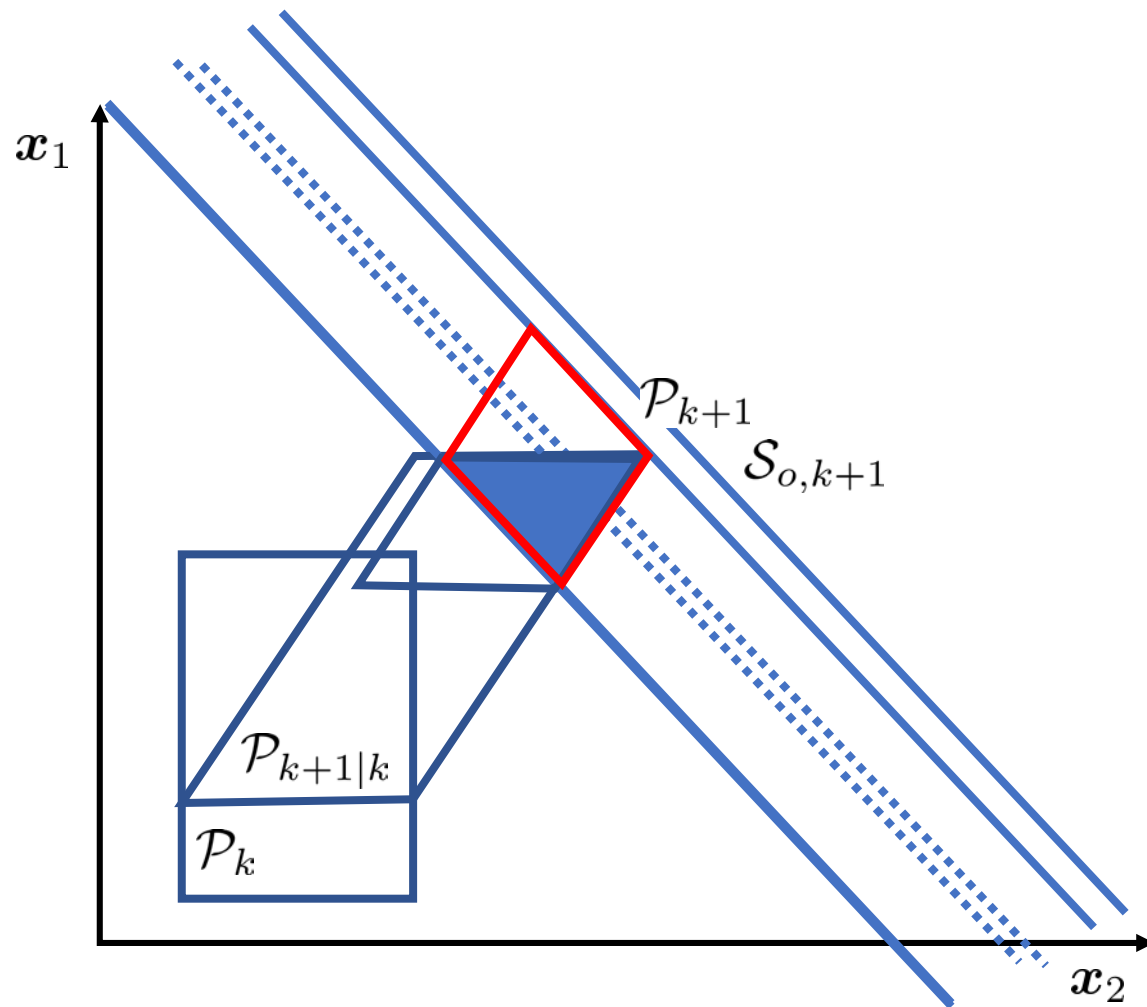
$$\mathcal{S}_k(\mathbf{p}_k, c_k) := \{\mathbf{x}_k \mid \mathbf{p}_k^\top \mathbf{x}_k - c_k \leq 1\}$$

$$\rightarrow \{\mathbf{x}_k \mid \mathbf{p}_k^\top (\mathbf{A}^{-1} \mathbf{x}_{k+1} - \mathbf{A}^{-1} \mathbf{B} \mathbf{u}) - c_k \leq 1\}$$

$$\rightarrow \{\mathbf{x}_{k+1} \mid \underbrace{\mathbf{p}_k^\top \mathbf{A}^{-1}} \mathbf{x}_{k+1} - \underbrace{(\mathbf{p}_k^\top \mathbf{A}^{-1} \mathbf{B} \mathbf{u} + c_k)} \leq 1\}$$



SSE with Parallelotopes (ROPO)



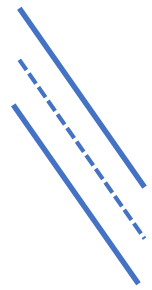
- Build the output set.
- Propagate the parallelotope.
- Find a parallelotope that enclose the intersection.

Recursive Optimal Parallelotopic Outbounding

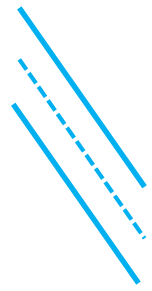
- ROPO algorithm is developed in three steps.
 - Step 1: Reduced Strip
 - Step 2: Reduced Parallelotope
 - Step 3: Choose the minimum volume parallelotope.

Vicino, A., & Zappa, G. (1996). **Sequential approximation of feasible parameter sets for identification with set membership uncertainty.** IEEE Transactions on Automatic Control, 41(6), 774-785.

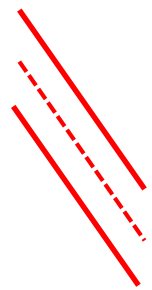
ROPO with Extremal Measurements (ROPOe)



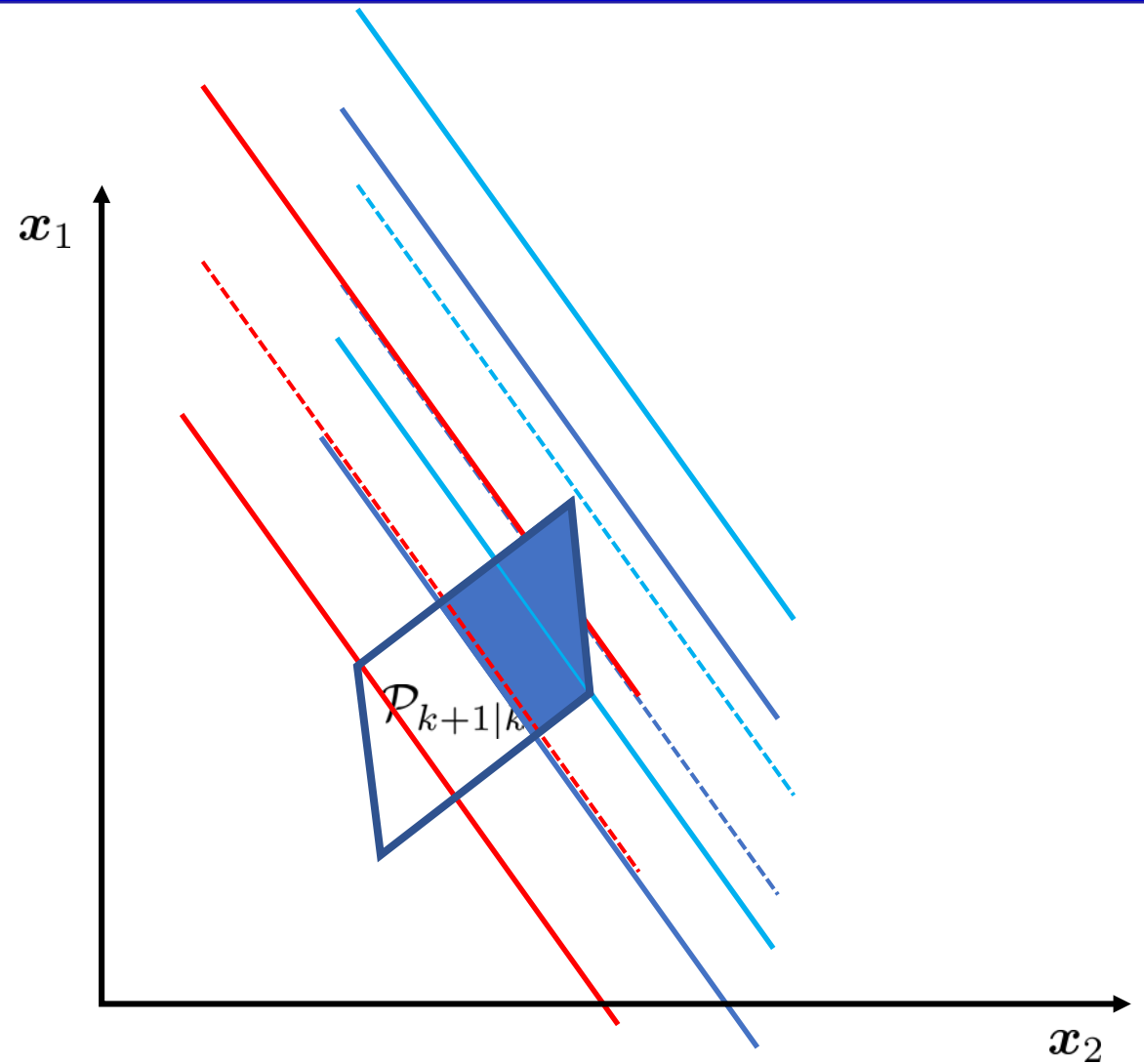
Current measurement



Maximal-error measurement



Minimal-error measurement



Valero, C. E., & Paulen, R. (2019). **Effective Recursive Set-membership State Estimation for Robust Linear MPC.** DYCOPS Conference, IFAC-Papers OnLine, 52(1), 486-491.

Problems

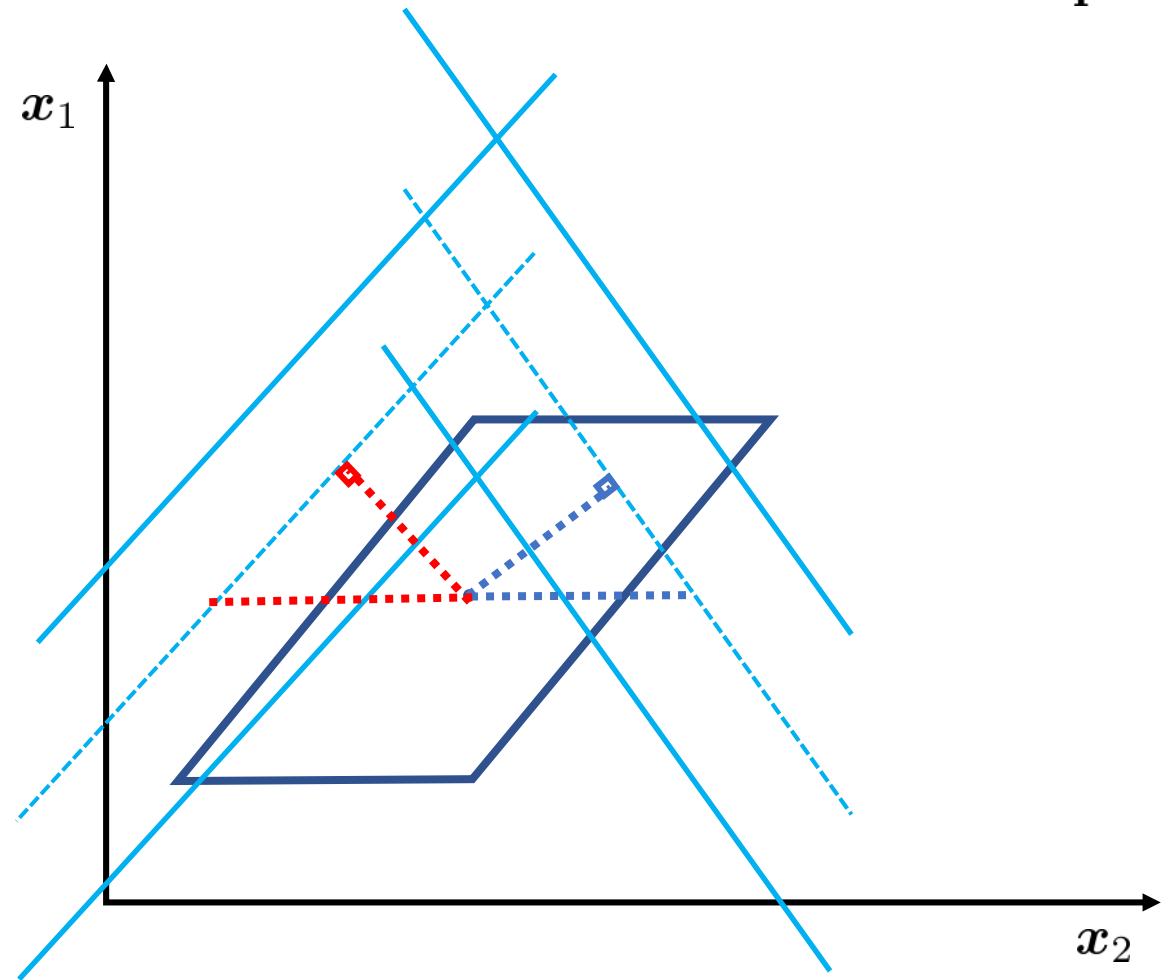
How to determine the best index criteria to find the minimum and maximum strip?

- Distance between a point and a plane.

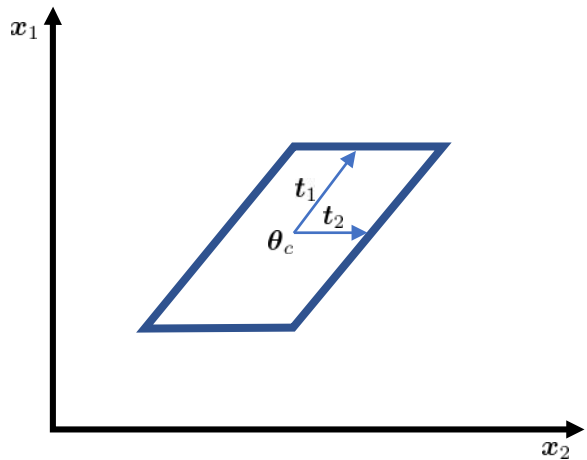
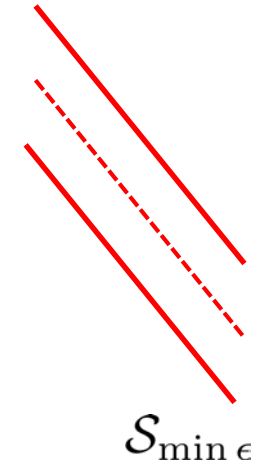
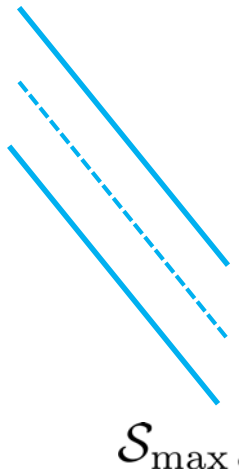
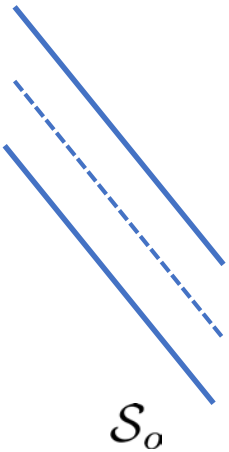
$$e := \frac{(\mathbf{p}^\top \boldsymbol{\theta}_c - c)}{\|\mathbf{p}\|_2}$$

- Projection between a point and a plane.

$$e := -C \frac{\mathbf{p}^\top \boldsymbol{\theta}_c}{\|\mathbf{p}^\top\|_2} \mathbf{p}$$



ROPO Extreme



- Keep strips with extremal measurements.
- Propagate the parallelotope and the extremal measurement strips through the linear dynamics.
- Perform estimation taking the new measurement information and the extremal measurements.
- Update the extremal-measurements strips based on the prediction error.

Valero, C. E., & Paulen, R. (2019). **Set-Theoretic State Estimation for Multi-output Systems using Block and Sequential Approaches**. In 2019 22nd International Conference on Process Control (PC19) (pp. 268-273). IEEE.

Simulation Studies

Case 1: SSE with Parallelotope

$$\mathbf{A} := \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad \mathbf{B} := \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \mathbf{E} = \boldsymbol{\eta} := \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \mathbf{C} := [1 \quad 0] \quad \mathbf{F} := 1$$

Constraints

$$-1 \leq u \leq 1$$

$$x_1 \geq 0$$

$$y_m = \mathbf{C}\mathbf{x} + \varepsilon$$

$$|\varepsilon| \leq 1$$

Initial conditions

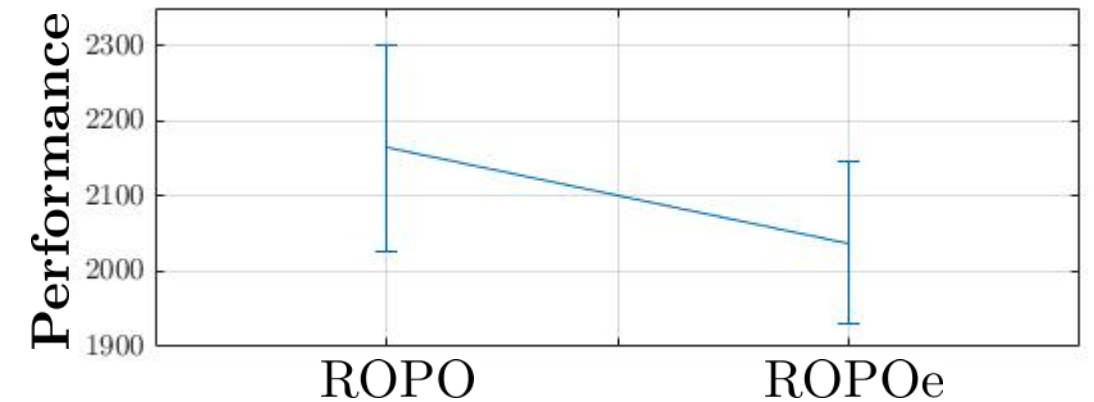
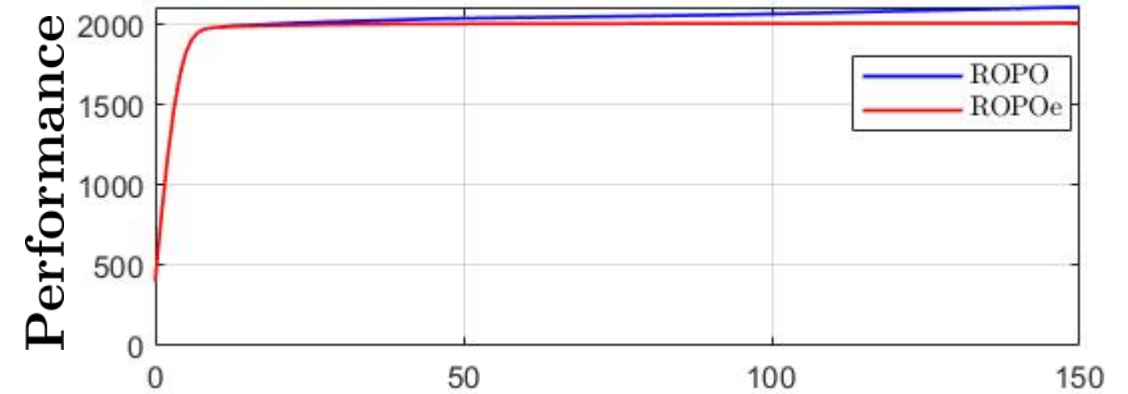
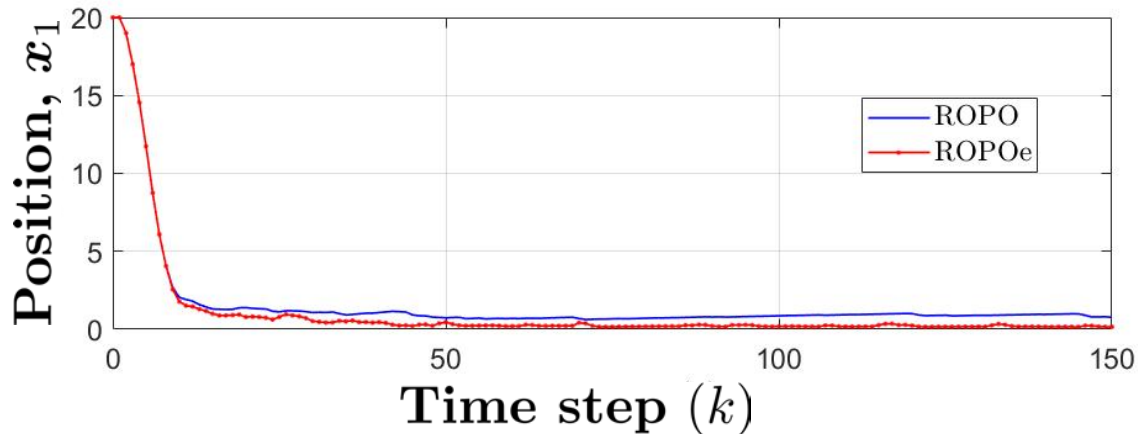
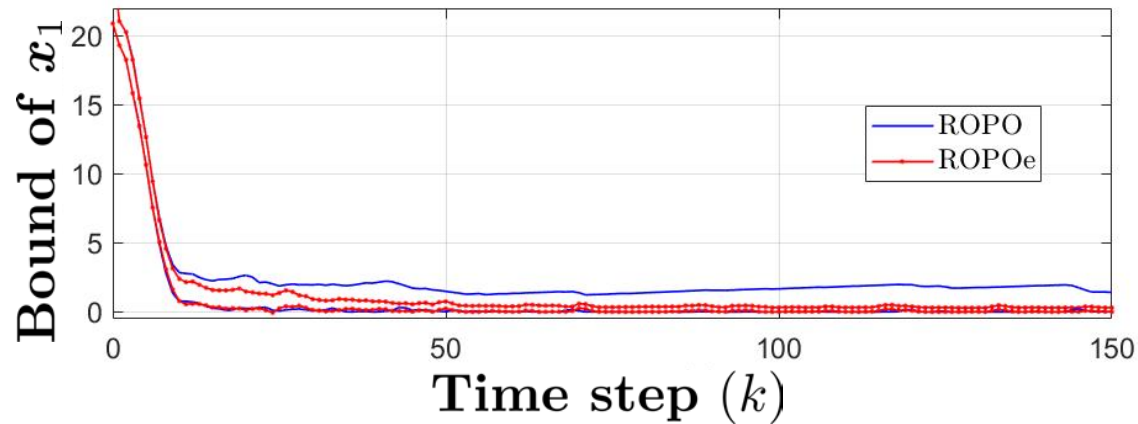
$$x_0 := \begin{bmatrix} 20 \\ 0 \end{bmatrix}$$

$$\mathcal{P}_0 := \left\{ \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \mathbf{v} + \begin{bmatrix} 21 \\ 1 \end{bmatrix} \mid \|\mathbf{v}\|_\infty \leq 1 \right\}$$

$$\mathbf{Q} = \mathbf{I}, R = 1 \times 10^{-6}, N_p = 10, N_u = N_p$$

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Simulation Studies



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Conclusions

- It is possible to enhance SSE taking into account past measurements of the system.

ROPO with extremal measurements(ROPOe)

- The propagation of one strip according to the dynamic of the plant.

$$\mathbf{p}_{k+1} := (\mathbf{p}_k^\top \mathbf{A}^{-1})^\top, \quad c_{k+1} := c_k + \mathbf{p}_k^\top \mathbf{A}^{-1} \mathbf{B} \mathbf{u}.$$

- The studied concepts apply to other bounded-error estimators.

Zonotopic Kalman Filter, Zonotopes, etc.