State Estimation for Linear Systems with Quadratic Outputs

Soulaimane Berkane¹ Dionysis Theodosis² Tarek Hamel ³ Dimos V. Dimarogonas ⁴

¹University of Quebec in Outaouais ²Technical University of Crete

³Côte d'Azur University ⁴KTH Royal Institute of Technology

American Control Conference, July 9-12, 2024









$$\dot{x} = Ax + Bu,$$

 $y_h = \frac{1}{2}x^{\top}C_hx + d_h^{\top}x, \ h = 1, \dots, q$

Without loss of generality, the matrices C_h are assumed to be symmetric.

$$\dot{x} = Ax + Bu,$$

 $y_h = \frac{1}{2}x^{\top}C_hx + d_h^{\top}x, \ h = 1, \dots, q$

Without loss of generality, the matrices C_h are assumed to be symmetric.

Its observability properties are affected by the input u

$$\dot{x} = Ax + Bu,$$

 $y_h = \frac{1}{2}x^{\top}C_hx + d_h^{\top}x, \ h = 1, \dots, q$

Without loss of generality, the matrices C_h are assumed to be symmetric.

Its observability properties are affected by the input u

Example (Observability depends on input)

Impossible to distinguish between initial conditions x_0 and $-x_0$ for

$$\begin{cases} \dot{x} = 0\\ y = x^2 \end{cases}$$

$$\dot{x} = Ax + Bu,$$

$$y_h = \frac{1}{2}x^{\top}C_hx + d_h^{\top}x, \ h = 1, \dots, q$$

Without loss of generality, the matrices C_h are assumed to be symmetric.

Robot localization from range measurements¹

$$\begin{cases} \dot{x} = u \\ y_h = \|x - a_h\|^2 - \|a_h\|^2 \end{cases}$$

¹T. Hamel and C. Samson, "Position estimation from direction or range measurements," *Automatica*, vol. 82, pp. 137–144, 2017.

11-11

$$\dot{x} = Ax + Bu,$$

$$y_h = \frac{1}{2}x^{\top}C_hx + d_h^{\top}x, \ h = 1, \dots, q$$

Without loss of generality, the matrices C_h are assumed to be symmetric.

We propose a systematic approach to **immerse** the original system into the higher order system

$$\dot{z} = \mathcal{A}(u)x + \mathcal{B}u,$$

 $y = \mathcal{C}z$

- Incorporating a minimum number of auxiliary states
- ► Uniform observability of the pair (A(u(t)), C) guarantees convergence of Kalman-type filters

Outline

Motivation

Single-Output Systems

Multiple-Output Systems

Conclusion and Future Work

Definition

- $\mathbb{M}_n \subset \mathbb{R}^{n \times n}$ denote the space of real $n \times n$ symmetric matrices
- The Lyapunov operator $L_A : \mathbb{M}_n \to \mathbb{M}_n$ is defined by

$$\mathbf{L}_{\mathcal{A}}(X) := X \mathcal{A} + \mathcal{A}^{\top} X$$

Definition

- $\mathbb{M}_n \subset \mathbb{R}^{n \times n}$ denote the space of real $n \times n$ symmetric matrices
- The Lyapunov operator $L_A : \mathbb{M}_n \to \mathbb{M}_n$ is defined by

$$\mathbf{L}_{A}(X) := XA + A^{\top}X$$

Observation: Any *quadratic form*, along the trajectories of the system, satisfies

$$\frac{1}{2}\frac{d}{dt}(x^{\top}Qx) = \frac{1}{2}x^{\top}\mathsf{L}_{\mathsf{A}}(\mathsf{Q})x + (QBu)^{\top}x, \qquad (1)$$

for some symmetric matrix $Q \in \mathbb{M}_n$.



Lemma

For any $Q \in \mathbb{M}_n$, there exists a minimum number *m* such that $m \le n(n+1)/2$ and

$$\mathbb{Q} := \operatorname{span}\{\mathsf{L}_{A}^{[0]}(Q), \mathsf{L}_{A}^{[1]}(Q), \dots, \mathsf{L}_{A}^{[m-1]}(Q)\}$$

is **L**_A-*invariant*.

Example

Consider the double-integrator system on \mathbb{R}^n

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = u \\ y = 0.5 \|x_1\|^2 \end{cases}$$

We have

$$\mathbf{L}_{A}^{[0]}(C) = \begin{bmatrix} I_{n} & 0\\ 0 & 0 \end{bmatrix}, \quad \mathbf{L}_{A}^{[1]}(C) = \begin{bmatrix} 0 & I_{n}\\ I_{n} & 0 \end{bmatrix}, \quad \mathbf{L}_{A}^{[2]}(C) = \begin{bmatrix} 0 & 0\\ 0 & 2I_{n} \end{bmatrix}$$

and finally, $L_A^{[3]}(C) = 0$. Hence

$$\operatorname{span}\{\mathsf{L}_A^{[0]}(Q),\mathsf{L}_A^{[1]}(Q),\mathsf{L}_A^{[2]}(Q)\}$$

is **L**_A-invariant.

Example

Consider the double-integrator system on \mathbb{R}^n

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = x_1 + 2x_2 + u \\ y = 0.5x_1^2 + 0.5x_2^2 \end{cases}$$

We have

$$\mathbf{L}_{A}^{[1]}(C) = \begin{bmatrix} 0 & 2\\ 2 & 4 \end{bmatrix}, \quad \mathbf{L}_{A}^{[2]}(C) = \begin{bmatrix} 4 & 8\\ 8 & 20 \end{bmatrix}$$

and finally, $\mathbf{L}_{A}^{[2]}(C) = 4(\mathbf{L}_{A}^{[0]}(C) + \mathbf{L}_{A}^{[1]}(C))$. Hence
 $\operatorname{span}\{\mathbf{L}_{A}^{[0]}(Q), \mathbf{L}_{A}^{[1]}(Q)\}$

is L_A -invariant. In this case, the number of auxiliary states needed is m = 2, which is strictly less than n(n + 1)/2 = 3.

Let us define the following auxiliary state variables

$$\xi_k := \frac{1}{2} x^\top \mathbf{L}_A^{[k]}(C) x, \quad k = 0, 1, \cdots, (m-1), \quad (2)$$

²By the previous lemma, there exist α_k s.th. $\mathbf{L}_A^{[m]}(Q) = \sum_{k=0}^{m-1} \alpha_k \mathbf{L}_A^{[k]}(Q)$ 10 / 21

Let us define the following auxiliary state variables

$$\xi_k := \frac{1}{2} x^\top \mathbf{L}_A^{[k]}(C) x, \quad k = 0, 1, \cdots, (m-1), \quad (2)$$

One has²

$$\dot{\xi}_{k} = \xi_{k+1} + (Bu)^{\top} \mathbf{L}_{A}^{[k]}(C)x, \ k = 0, \dots, (m-2)$$
$$\dot{\xi}_{m-1} = \sum_{k=0}^{m-1} \alpha_{k} \xi_{k} + (Bu)^{\top} \mathbf{L}_{A}^{[m-1]}(C)x.$$
(3)

²By the previous lemma, there exist α_k s.th. $\mathbf{L}_A^{[m]}(Q) = \sum_{k=0}^{m-1} \alpha_k \mathbf{L}_A^{[k]}(Q)$ 10 / 21

$$\begin{aligned} \text{Single-Output Systems} \\ \text{Define } z &:= \begin{bmatrix} \xi_{m-1} & \cdots & \xi_1 & \xi_0 & x^\top \end{bmatrix}^\top \in \mathbb{R}^{m+n} : \\ \hline \dot{z} &= \mathcal{A}(u)x + \mathcal{B}u, \\ y &= \mathcal{C}z \end{aligned}$$
$$\mathcal{A}(u) &= \begin{bmatrix} \alpha_{m-1} & \alpha_{m-2} & \cdots & \alpha_1 & \alpha_0 & (Bu)^\top \mathbf{L}_A^{[m-1]}(\mathcal{C}) \\ 1 & 0 & \cdots & 0 & 0 & (Bu)^\top \mathbf{L}_A^{[m-2]}(\mathcal{C}) \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & 0 & (Bu)^\top \mathbf{L}_A^{[0]}(\mathcal{C}) \\ \hline 0 & 0 & \cdots & 0 & 0 & | & \mathcal{A} \end{bmatrix}, \\ \mathcal{B} &:= \begin{bmatrix} 0 & 0 & \cdots & | & \mathcal{B}^\top \end{bmatrix}^\top, \quad \mathcal{C} := \begin{bmatrix} 0 & 0 & \cdots & 0 & 1 & | & d^\top \end{aligned}$$

<u>Uniform observability has been discussed</u> in³ when $\alpha_k = 0$

³D. Theodosis, S. Berkane, and D. V. Dimarogonas, "State estimation for a class of linear systems with quadratic output," *IFAC-PapersOnLine*, vol. 54, no. 9, pp. 261–266, 2021.

The naive extension of the single-output procedure to multiple-outputs does not guarantee the minimum number of auxiliary states



$$\begin{split} \dim(\mathbb{C}_1 + \mathbb{C}_2) &= \dim(\mathbb{C}_1) + \dim(\mathbb{C}_2) - \dim(\mathbb{C}_1 \cap \mathbb{C}_2) \\ &\leq \dim(\mathbb{C}_1) + \dim(\mathbb{C}_2) \end{split}$$

$$y_h = \frac{1}{2} x^\top C_h x + d_h^\top x$$

(4)

$$y_{h} = \frac{1}{2} x^{\top} C_{h} x + d_{h}^{\top} x$$

= $\frac{1}{2} \operatorname{vec}(C_{h})^{\top} (x \otimes x) + d_{h}^{\top} x$ (4)

$$y_{h} = \frac{1}{2} x^{\top} C_{h} x + d_{h}^{\top} x$$

$$= \frac{1}{2} \operatorname{vec}(C_{h})^{\top} (x \otimes x) + d_{h}^{\top} x$$

$$= \frac{1}{2} \operatorname{vech}(C_{h})^{\top} D_{n}^{\top} (x \otimes x) + d_{h}^{\top} x.$$
 (4)

⁴For $A \in \mathbb{M}_n$, $\operatorname{vech}(A)$ is obtained by vectorizing only the lower triangular part of A. One has $D_n \operatorname{vech}(A) = \operatorname{vec}(A)$ where D_n is the *duplication matrix* 3 / 21

$$y_{h} = \frac{1}{2} x^{\top} C_{h} x + d_{h}^{\top} x$$

$$= \frac{1}{2} \operatorname{vec}(C_{h})^{\top} (x \otimes x) + d_{h}^{\top} x$$

$$= \frac{1}{2} \operatorname{vech}(C_{h})^{\top} D_{n}^{\top} (x \otimes x) + d_{h}^{\top} x.$$

$$= \frac{1}{2} \operatorname{vech}(C_{h})^{\top} x^{[2]} + d_{h}^{\top} x.$$
(4)

⁴For $A \in \mathbb{M}_n$, $\operatorname{vech}(A)$ is obtained by vectorizing only the lower triangular part of A. One has $D_n \operatorname{vech}(A) = \operatorname{vec}(A)$ where D_n is the *duplication matrix* 3 / 21

$$y_{h} = \frac{1}{2} x^{\top} C_{h} x + d_{h}^{\top} x$$

$$= \frac{1}{2} \operatorname{vec}(C_{h})^{\top} (x \otimes x) + d_{h}^{\top} x$$

$$= \frac{1}{2} \operatorname{vech}(C_{h})^{\top} D_{n}^{\top} (x \otimes x) + d_{h}^{\top} x.$$

$$= \frac{1}{2} \operatorname{vech}(C_{h})^{\top} x^{[2]} + d_{h}^{\top} x.$$
(4)

Example

$$(x \otimes x) = \begin{bmatrix} x_1^2 \\ x_1 x_2 \\ x_2 x_1 \\ x_2^2 \end{bmatrix}, \quad D_2 \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow x^{[2]} = \begin{bmatrix} x_1^2 \\ 2x_1 x_2 \\ x_2^2 \end{bmatrix}$$

⁴For $A \in \mathbb{M}_n$, $\operatorname{vech}(A)$ is obtained by vectorizing only the lower triangular part of A. One has $D_n \operatorname{vech}(A) = \operatorname{vec}(A)$ where D_n is the *duplication matrix* 3 / 21

The output vector $y := [y_1 \cdots y_q]^\top$ can be written as

$$y = \frac{1}{2} \begin{bmatrix} \operatorname{vech}(C_1)^\top \\ \vdots \\ \operatorname{vech}(C_q)^\top \end{bmatrix} x^{[2]} + \begin{bmatrix} d_1^\top \\ \vdots \\ d_q^\top \end{bmatrix} x =: \bar{C} x^{[2]} + Dx.$$
(5)

Lemma

Along the trajectories of the system, one has

$$\frac{d}{dt}x^{[2]} = \bar{A}x^{[2]} + \bar{U}(t)x.$$
(6)

with $\bar{A} := D_n^\top (A \oplus A) (D_n^+)^\top$ and $\bar{U}(t) := D_n^\top (Bu(t) \oplus Bu(t)).$

The extended state z[⊤] := [(x^[2])[⊤] x[⊤]] leads to an LTV system. The number of added auxiliary states is n(n + 1)/2.

- In this work, we propose a procedure to add the minimum number of auxiliary states to form an LTV system.
- Rank factorization⁵

$$\frac{1}{2} \begin{bmatrix} \operatorname{vech}(C_1)^\top \\ \vdots \\ \operatorname{vech}(C_q)^\top \end{bmatrix} = \bar{C} = F\bar{L}_0, \quad \operatorname{rank}(\bar{L}_0) = \operatorname{rank}(\bar{C}) = p_0$$

• Let $\xi_0 \in \mathbb{R}^{p_0}$ such that

$$\xi_0 := \bar{L}_0 x^{[2]} = \begin{bmatrix} \frac{1}{2} x^\top L_1 x & \cdots & \frac{1}{2} x^\top L_{\rho_0} x \end{bmatrix}^\top.$$
(7)

We have⁶

$$\dot{\xi}_0 = \bar{L}_0 \bar{A} x^{[2]} + \bar{L}_0 \bar{U}(t) x.$$
 (8)

⁵ \bar{L}_0 is full row rank and F is full column rank.

 $^{{}^6\}dot{\xi_0}$ will be linear in ξ_0 and x if and only if $\mathrm{span}\{L_1,\cdots,L_{p_0}\}$ is L_A- invariant

Assume
$$p_{1} := \operatorname{rank} \begin{bmatrix} \bar{L}_{0} \\ \bar{L}_{0}\bar{A} \end{bmatrix} - p_{0} \neq 0 \tag{9}$$

• There exist $P_0, M_0^{(0)}, \overline{L}_1$ such that⁷

$$P_0 \bar{L}_0 \bar{A} = \begin{bmatrix} \bar{L}_1 \\ M_0^{(0)} P_0 \bar{L}_0 \end{bmatrix}.$$
 (10)

• If we let $\xi_1 := \overline{L}_1 x^{[2]}$, it follows that

$$P_0 \dot{\xi}_0 = \begin{bmatrix} 0\\ M_0^{(0)} \end{bmatrix} P_0 \xi_0 + \begin{bmatrix} I_{p_1}\\ 0 \end{bmatrix} \xi_1 + P_0 \bar{L}_0 \bar{U} x.$$
(11)

 $^{7}P_{k}$ is a permutation matrix

Algorithm 1 Computation of the matrices $\bar{L}_k, P_k, M_i^{(k)}$

Require: Output matrix \overline{C} and matrix \overline{A}

- 1: Compute $p_0 := \operatorname{rank}(\bar{C})$
- 2: Rank factorize $\bar{C} = F\bar{L}_0$ with rank $(\bar{L}_0) = p_0$
- 3: for $k \in \{1, 2, \dots\}$ do
- 4: Calculate

$$p_k := \operatorname{rank} \begin{bmatrix} \bar{L}_0 \\ \vdots \\ \bar{L}_{k-1} \\ \bar{L}_{k-1} \bar{A} \end{bmatrix} - \sum_{i=0}^{k-1} p_i$$
(30)

5: Find a permutation matrix $P_{k-1} \in \mathbb{R}^{p_{k-1} \times p_{k-1}}$, matrices $M_i^{(k-1)} \in \mathbb{R}^{(p_{k-1}-p_k) \times p_i}$ and a full row matrix $\bar{L}_k \in \mathbb{R}^{p_k \times n(n+1)/2}$ satisfying¹

$$P_{k-1}\bar{L}_{k-1}\bar{A} = \begin{bmatrix} \bar{L}_k \\ \sum_{i=0}^{k-1} M_i^{(k-1)} P_i \bar{L}_i \end{bmatrix}$$
(31)

6: **if** $p_k = 0$ **then** 7: Define m = k8: **return** the matrices $\bar{L}_{k-1}, P_{k-1}, M_{i-1}^{(k-1)}$ for all $1 \le i \le k \le m$. 9: **end if** 10: **end for**

We define the extended state vector

$$z := \begin{bmatrix} P_{m-1}\xi_{m-1} \\ \vdots \\ P_{0}\xi_{0} \\ x \end{bmatrix} := \begin{bmatrix} P_{m-1}\bar{L}_{m-1}x^{[2]} \\ \vdots \\ P_{0}\bar{L}_{0}x^{[2]} \\ x \end{bmatrix} \in \mathbb{R}^{n+\sum_{k=0}^{m-1}p_{k}}, \quad (12)$$

the dynamics of z are an LTV system with matrices

$$\mathcal{A}(u) = \begin{bmatrix} \bar{M}_{m-1}^{(m-1)} & \cdots & \cdots & \bar{M}_{0}^{(m-1)} & P_{m-1}\bar{L}_{m-1}\bar{U} \\ \bar{P}_{m-1} & \cdots & \cdots & \bar{M}_{0}^{(m-2)} & P_{m-2}\bar{L}_{m-2}\bar{U} \\ \vdots & \ddots & \vdots & \vdots & & \vdots \\ 0 & \cdots & \bar{P}_{1} & M_{0}^{(0)} & P_{0}\bar{L}_{0}\bar{U} \\ \hline 0 & & & A \end{bmatrix}, \quad (13)$$
$$\mathcal{B} = \begin{bmatrix} 0 \\ \bar{B} \end{bmatrix}, \ \mathcal{C} = \begin{bmatrix} 0 & \cdots & 0 & FP_{0}^{\top} & D \end{bmatrix}, \quad (14)$$

Example

Consider the system

$$\begin{cases} \dot{p} &= v_a + v_w \\ \dot{v}_a &= u(t) \\ \dot{v}_w &= 0 \end{cases} \qquad \begin{cases} y_1 &= 0.5 \|p\|^2 \\ y_2 &= 0.5 \|v_a\|^2 \end{cases}$$

The proposed algorithm returns m = 3 with

$$\xi_{2} = 2 \|v_{a} + v_{w}\|^{2}$$

$$\xi_{1} = 2p^{\top}(v_{a} + v_{w})$$

$$\xi_{0} = \begin{bmatrix} \|p\|^{2} \\ \|v_{a}\|^{2} \end{bmatrix}$$

These auxiliary states allow to bring the system dynamics to an LTV form where a Kalman-type estimator can be applied.

Example

Consider the system

$$\begin{cases} \dot{p} &= v_a + v_w \\ \dot{v}_a &= u(t) \\ \dot{v}_w &= 0 \end{cases} \qquad \begin{cases} y_1 &= 0.5 \|p\|^2 \\ y_2 &= 0.5 \|v_a\|^2 \end{cases}$$



19/21

Conclusion and Future Work

Summary:

- An immersion-type technique that transforms LTI systems with quadratic outputs to an LTV system with linear output
- The approach guarantees that only the minimum number of auxiliary states is introduced
- The resultant systems observability is tightly related to the richness of the input signal

Future work:

- Extend the approach to LTV systems with quadratic outputs
- Extend the approach to polynomial outputs
- Design reduced-order observers to estimate only the original state

Thank you

Questions?