

# The shortest experiment for linear system identification

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**Joint work with Kanat Camlibel and Paolo Rapisarda**

- 1 The problem of experiment design
  - ▶ fundamental lemma
  - ▶ online experiment design
- 2 Informativity for system identification
- 3 The shortest experiment

# The problem of experiment design

# Experiment design

True system:

$$\begin{aligned}x(t+1) &= A_{\text{true}}x(t) + B_{\text{true}}u(t) \\y(t) &= C_{\text{true}}x(t) + D_{\text{true}}u(t)\end{aligned}$$

$$\begin{bmatrix} A_{\text{true}} & B_{\text{true}} \\ C_{\text{true}} & D_{\text{true}} \end{bmatrix} \in \mathbb{R}^{(n_{\text{true}}+p) \times (n_{\text{true}}+m)} \text{ and } n_{\text{true}} \text{ are unknown}$$

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Observability matrix and lag:

For  $k \geq 0$  we define

$$\Omega_k = \begin{cases} 0_{0,n} & \text{if } k = 0 \\ \begin{bmatrix} C_{\text{true}} \\ C_{\text{true}}A_{\text{true}} \\ \vdots \\ C_{\text{true}}A_{\text{true}}^{k-1} \end{bmatrix} & \text{if } k \geq 1 \end{cases}$$

The **lag** is defined as the smallest integer  $\ell \geq 0$  such that  $\text{rank } \Omega_\ell = \text{rank } \Omega_{\ell+1}$  and denoted by  $\ell_{\text{true}} = \ell(C_{\text{true}}, A_{\text{true}}) \leq n_{\text{true}}$ .

# Experiment design

True system:

$$\begin{aligned}x(t+1) &= A_{\text{true}}x(t) + B_{\text{true}}u(t) \\ y(t) &= C_{\text{true}}x(t) + D_{\text{true}}u(t)\end{aligned}\tag{1}$$

$$\begin{bmatrix} A_{\text{true}} & B_{\text{true}} \\ C_{\text{true}} & D_{\text{true}} \end{bmatrix} \in \mathbb{R}^{(n_{\text{true}}+p) \times (n_{\text{true}}+m)} \text{ and } n_{\text{true}} \text{ are } \mathbf{unknown}$$

Prior knowledge: (1) is **controllable** and **observable**,  $\ell_{\text{true}} \leq L$  and  $n_{\text{true}} \leq N$

Fundamental question: How to find  $T \in \mathbb{N}$  and

$$u_{[0, T-1]} := [u(0) \quad u(1) \quad \cdots \quad u(T-1)]$$

such that the resulting **data**  $(u_{[0, T-1]}, y_{[0, T-1]})$  **enable system identification?**

I.e., such that we can **identify**  $n_{\text{true}}$  and matrices  $A, B, C$  and  $D$  satisfying

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} S & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} A_{\text{true}} & B_{\text{true}} \\ C_{\text{true}} & D_{\text{true}} \end{bmatrix} \begin{bmatrix} S^{-1} & 0 \\ 0 & I \end{bmatrix} \text{ for some invertible } S$$

### A note on persistency of excitation

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#### Abstract

We prove that if a component of the response signal of a controllable linear time-invariant system is persistently exciting of sufficiently high order, then the windows of the signal span the full system behavior. This is then applied to obtain conditions

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**Definition:** The input  $u_{[0, T-1]}$  is called **persistently exciting** of order  $k$  if

$$\text{rank } H_k(u_{[0, T-1]}) = \text{rank} \begin{bmatrix} u(0) & u(1) & \cdots & u(T-k) \\ \vdots & \vdots & & \vdots \\ u(k-1) & u(k) & \cdots & u(T-1) \end{bmatrix} = km$$

# A partial answer

## fundamental lemma

### Possible solution:

- Choose  $T := (N + L + 1)m + N + L$
- Design  $u_{[0, T-1]}$  to be **persistently exciting of order  $N + L + 1$**
- Then by the **fundamental lemma**,

$$\text{rank} \begin{bmatrix} H_{L+1}(u_{[0, T-1]}) \\ \hline H_{L+1}(y_{[0, T-1]}) \end{bmatrix} = \begin{bmatrix} u(0) & \cdots & u(T-L-1) \\ \vdots & & \vdots \\ u(L) & \cdots & u(T-1) \\ \hline y(0) & \cdots & y(T-L-1) \\ \vdots & & \vdots \\ y(L) & \cdots & y(T-1) \end{bmatrix} = (L+1)m + n_{\text{true}}$$

- Apply **subspace identification** to obtain  $A, B, C$  and  $D$

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We will now consider a simple example...

# A partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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Hence,  $n_{\text{true}} = 2$  and  $\ell_{\text{true}} = 1$ . We take  $N = L = 2$ .

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Define  $T = 9$  and  $u_{[0,8]} := [1 \ 1 \ 0 \ 0 \ 0 \ 0 \ 1 \ 0 \ 0]$  (PE of order 5)

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$$\text{rank} \begin{bmatrix} H_3(u_{[0,8]}) \\ \hline H_3(y_{[0,8]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ \hline -1 & 0 & 0 & 1 & 1 & 2 & 3 \\ 2 & 2 & 1 & 1 & 2 & 3 & 7 \\ 0 & 0 & 1 & 1 & 2 & 3 & 5 \\ 2 & 1 & 1 & 2 & 3 & 7 & 9 \\ 0 & 1 & 1 & 2 & 3 & 5 & 9 \\ 1 & 1 & 2 & 3 & 7 & 9 & 14 \end{bmatrix} = 5 \quad \Longrightarrow \quad n_{\text{true}} = 2$$



# A partial answer

## fundamental lemma

### Possible solution:

- Choose  $T := (N + L + 1)m + N + L$
- Design  $u_{[0, T-1]}$  to be **persistently exciting of order  $N + L + 1$**
- Then by the **fundamental lemma**,

$$\text{rank} \begin{bmatrix} H_{L+1}(u_{[0, T-1]}) \\ H_{L+1}(y_{[0, T-1]}) \end{bmatrix} = \begin{bmatrix} u(0) & \cdots & u(T - L - 1) \\ \vdots & & \vdots \\ u(L) & \cdots & u(T - 1) \\ y(0) & \cdots & y(T - L - 1) \\ \vdots & & \vdots \\ y(L) & \cdots & y(T - 1) \end{bmatrix} = (L + 1)m + n_{\text{true}}$$

- Apply **subspace identification** to obtain  $A, B, C$  and  $D$

**Question:** Is this the **smallest possible  $T$** ?

**Answer:** **no!**

## Beyond Persistent Excitation: Online Experiment Design for Data-Driven Modeling and Control

Henk J. van Waarde<sup>1b</sup>

**Abstract**—This letter presents a new experiment design method for data-driven modeling and control. The idea is to select inputs *online* (using past input/output data), leading to desirable rank properties of data Hankel matrices. In

rank property is important, since it guarantees that *all* trajectories of the system can be parameterized in terms of the measured trajectory. Essentially, the Hankel matrix of measured inputs and outputs serves as a non-parametric model of

### Possible solution:

- Design the input  $u(t)$  **online** based on  $(u_{[0,t-1]}, y_{[0,t-1]})$
- $T$  is not specified a priori, but procedure terminates after  $T = (L + 1)m + n_{\text{true}} + L$  steps
- Apply **subspace identification** to obtain  $A, B, C$  and  $D$

We again consider an example...

# Another partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$ , measure  $y_{[0,2]} = \begin{bmatrix} -1 & 0 & 0 \\ 2 & 0 & 0 \end{bmatrix}$ ;

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$$\text{rank} \begin{bmatrix} H_3(u_{[0,2]}) \\ \hline H_2(y_{[0,1]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 \\ 0 \\ 0 \\ \hline -1 \\ 2 \\ 0 \\ 0 \end{bmatrix} = 1$$

# Another partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$ , measure  $y_{[0,2]} = \begin{bmatrix} -1 & 0 & 0 \\ 2 & 0 & 0 \end{bmatrix}$ ;

$$\text{rank} \begin{bmatrix} H_3(u_{[0,3]}) \\ \hline H_2(y_{[0,2]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & \mathbf{u(3)} \\ \hline -1 & 0 \\ 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} = 2 \text{ for } \mathbf{u(3)} = 1$$

$$\text{Measure } y(3) = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$$

# Another partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$  and design the rest of the inputs **online**

$$\text{rank} \begin{bmatrix} H_3(u_{[0,4]}) \\ \hline H_2(y_{[0,3]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & u(4) \\ \hline -1 & 0 & 0 \\ 2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 2 \end{bmatrix} = 3 \text{ for any } u(4)$$

Take  $u(4) = 0$

# Another partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$  and design the rest of the inputs **online**

$$\text{rank} \begin{bmatrix} H_3(u_{[0,5]}) \\ \hline H_2(y_{[0,4]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & u(5) \\ \hline -1 & 0 & 0 & 0 \\ 2 & 0 & 0 & 2 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 1 \end{bmatrix} = 4 \text{ for any } u(5)$$

Take  $u(5) = 0$

# Another partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$  and design the rest of the inputs **online**

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$$\text{rank} \begin{bmatrix} H_3(u_{[0,6]}) \\ \hline H_2(y_{[0,5]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & u(6) \\ \hline -1 & 0 & 0 & 0 & 0 \\ 2 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 2 & 1 & 1 \end{bmatrix} = 5 \text{ for any } u(6)$$

So we take  $u(6) = 0$ .

# Another partial answer

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$  and design the rest of the inputs **online**

$$\text{rank} \begin{bmatrix} H_3(u_{[0,7]}) \\ H_2(y_{[0,6]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & u(7) \\ \hline -1 & 0 & 0 & 0 & 0 & 1 \\ 2 & 0 & 0 & 2 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 2 & 1 & 1 & 2 \end{bmatrix} = 5 \neq 6 \text{ for any } u(7)$$

So we do not apply  $u(7)$  and **stop the procedure**.



# Another partial answer

example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Define  $u_{[0,2]} = [1 \ 0 \ 0] \neq 0$  and design the rest of the inputs **online**

It follows that

$$\text{rank} \begin{bmatrix} H_3(u_{[0,6]}) \\ \hline H_3(y_{[0,6]}) \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ \hline -1 & 0 & 0 & 0 & 0 \\ 2 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 2 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 2 & 1 & 1 & 2 \end{bmatrix} = (L+1)m + n_{\text{true}} = 5 \implies n_{\text{true}} = 2$$

**Reduction # of samples:** from  $T = 9$  to  $T = 7$

## Beyond Persistent Excitation: Online Experiment Design for Data-Driven Modeling and Control

Henk J. van Waarde<sup>1b</sup>

**Abstract**—This letter presents a new experiment design method for data-driven modeling and control. The idea is to select inputs *online* (using past input/output data), leading to desirable rank properties of data Hankel matrices. In

rank property is important, since it guarantees that *all* trajectories of the system can be parameterized in terms of the measured trajectory. Essentially, the Hankel matrix of measured inputs and outputs serves as a non-parametric model of

### Possible solution:

- Design the input  $u(t)$  **online** based on  $(u_{[0,t-1]}, y_{[0,t-1]})$
- $T$  is not specified a priori, but procedure terminates after  $T = (L + 1)m + n_{\text{true}} + L$  steps
- Apply **subspace identification** to obtain  $A, B, C$  and  $D$

**Question:** Is this the **smallest possible**  $T$ ?

**Answer:** **it's a secret!**

# Informativity for system identification

## Beyond the fundamental lemma: from finite time series to linear system

M. Kanat Camlibel<sup>1</sup> and Paolo Rapisarda<sup>2</sup>

<sup>1</sup>Bernoulli Institute, University of Groningen

<sup>2</sup>School of Electronics and Computer Science, University of Southampton

### Abstract

We state necessary and sufficient conditions to uniquely identify (modulo state isomorphism) a linear time-invariant minimal input-state-output system from finite input-output data and upper- and lower bounds on lag and state space dimension.

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**Data:** Let  $(u_{[0, T-1]}, y_{[0, T-1]})$  be generated by the true system (**no assumptions on the input for now!**)

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**Question:** Under what conditions on  $(u_{[0, T-1]}, y_{[0, T-1]})$  can we **uniquely identify** the true system (up to state-space transformations)?

# System classes

$$\begin{aligned}x(t+1) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned} \quad \longleftrightarrow \quad \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathbb{R}^{(n+p) \times (n+m)}$$

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$$\mathcal{S} = \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathbb{R}^{(n+p) \times (n+m)} \mid n \geq 0 \right\} \quad \text{systems with } m \text{ inputs and } p \text{ outputs}$$

$$\mathcal{O} = \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{S} \mid (C, A) \text{ is } \mathbf{observable} \right\} \quad \text{observable systems}$$

$$\mathcal{M} = \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{O} \mid (A, B) \text{ is } \mathbf{controllable} \right\} \quad \text{minimal systems}$$

---

$$\mathcal{S}(n) = \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{S} \mid A \in \mathbb{R}^{n \times n} \right\} \quad \text{systems with } n \text{ states}$$

$$\mathcal{S}(\ell, n) = \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{S}(n) \mid \ell(C, A) = \ell \right\} \quad \text{systems with lag } \ell \text{ and } n \text{ states}$$

# Explaining systems

**Definition:** A system  $\begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathbb{R}^{(n+p) \times (n+m)}$  **explains** the data  $(u_{[0, T-1]}, y_{[0, T-1]})$  if

$$\begin{bmatrix} \mathbf{x}_{[1, T]} \\ y_{[0, T-1]} \end{bmatrix} = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} \mathbf{x}_{[0, T-1]} \\ u_{[0, T-1]} \end{bmatrix}$$

for some  $\mathbf{x}_{[0, T]} \in \mathbb{R}^{n \times (T+1)}$ .

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$\mathcal{E} = \{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{S} \mid \begin{bmatrix} A & B \\ C & D \end{bmatrix} \text{ explains the data } (u_{[0, T-1]}, y_{[0, T-1]}) \}$  explaining systems

$\mathcal{E}(n) = \mathcal{E} \cap \mathcal{S}(n)$  explaining systems with  $n$  states

$\mathcal{E}(\ell, n) = \mathcal{E} \cap \mathcal{S}(\ell, n)$  explaining systems with lag  $\ell$  and  $n$  states

---

**True system:**

$$\begin{bmatrix} A_{\text{true}} & B_{\text{true}} \\ C_{\text{true}} & D_{\text{true}} \end{bmatrix} \in \mathcal{E}(\ell_{\text{true}}, n_{\text{true}}) \subseteq \mathcal{E}(n_{\text{true}}) \subseteq \mathcal{E}$$

# System identification

Prior knowledge:  $\mathcal{S}_{\text{pk}} \subseteq \mathcal{S}$  with  $\begin{bmatrix} A_{\text{true}} & B_{\text{true}} \\ C_{\text{true}} & D_{\text{true}} \end{bmatrix} \in \mathcal{S}_{\text{pk}}$

---

Upper bounds on the lag and state dimension:

- Recall that

$$\ell_{\text{true}} \leq L \quad \text{and} \quad n_{\text{true}} \leq N$$

- Define

$$\mathcal{S}_{L,N} := \left\{ \begin{bmatrix} A & B \\ C & D \end{bmatrix} \in \mathcal{S}(\ell, n) \mid \ell \leq L \text{ and } n \leq N \right\}$$

- Our prior knowledge is thus:

$$\mathcal{S}_{\text{pk}} = \mathcal{S}_{L,N} \cap \mathcal{M}$$

---

**Definition:** The data  $(u_{[0,T-1]}, y_{[0,T-1]})$  are **informative for SysId** if

- $\mathcal{E} \cap \mathcal{S}_{\text{pk}} = \mathcal{E}(n_{\text{true}}) \cap \mathcal{S}_{\text{pk}}$  (data determine state dimension)
- Any pair of systems in  $\mathcal{E} \cap \mathcal{S}_{\text{pk}}$  is isomorphic

# Necessary and sufficient conditions

$\ell_{\min} = \min\{\ell \geq 0 \mid \exists n \geq 0 \text{ s.t. } \mathcal{E}(\ell, n) \neq \emptyset\}$       minimum lag to explain the data

$n_{\min} = \min\{n \geq 0 \mid \mathcal{E}(n) \neq \emptyset\}$       minimum state dimension to explain the data

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**Theorem:**  $\mathcal{E}(\ell, n) \neq \emptyset \implies n - \ell \geq n_{\min} - \ell_{\min} \implies \ell \leq n - n_{\min} + \ell_{\min}$

---

**Observation:**  $L_d := N - n_{\min} + \ell_{\min}$       data-guided bound on lag  
 $L_a := \min(L, L_d)$       actual upper bound

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**Theorem** (Camlibel and Rapisarda, 2024): The data  $(u_{[0, T-1]}, y_{[0, T-1]})$  are informative for SysId **if and only if**

$$T \geq L_a + (L_a + 1)m + n_{\min}$$

and

$$\text{rank} \begin{bmatrix} H_{L_a+1}(u_{[0, T-1]}) \\ H_{L_a+1}(y_{[0, T-1]}) \end{bmatrix} = (L_a + 1)m + n_{\min}.$$

Moreover, if these conditions are satisfied then  $\ell_{\text{true}} = \ell_{\min}$  and  $n_{\text{true}} = n_{\min}$ .



# The shortest experiment

# Necessary conditions for informativity

**Recall:**  $L_a := \min(L, N - n_{\min} + \ell_{\min})$

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**Theorem:** The data  $(u_{[0, T-1]}, y_{[0, T-1]})$  are informative for SysId **if and only if**

$$T \geq L_a + (L_a + 1)m + n_{\min} \quad \text{and} \quad \text{rank} \begin{bmatrix} H_{L_a+1}(u_{[0, T-1]}) \\ H_{L_a+1}(y_{[0, T-1]}) \end{bmatrix} = (L_a + 1)m + n_{\min}.$$

Moreover, if these conditions are satisfied, then  $\ell_{\text{true}} = \ell_{\min}$  and  $n_{\text{true}} = n_{\min}$ .

---

**Observation:** The **shortest** possible informative data length is

$$T := L + (L + 1)m + n_{\text{true}} \quad \text{where} \quad L := \min(L, N - n_{\text{true}} + \ell_{\text{true}})$$

---

**Question:** Is it possible to **generate** informative data  $(u_{[0, T-1]}, y_{[0, T-1]})$ , i.e.,

$$\text{rank} \begin{bmatrix} H_{L+1}(u_{[0, T-1]}) \\ H_{L+1}(y_{[0, T-1]}) \end{bmatrix} = (L + 1)m + n_{\text{true}}$$

without knowing  $\ell_{\text{true}}$  and  $n_{\text{true}}$ ?

# Preparation

For the data  $(u_{[0,t-1]}, y_{[0,t-1]})$ , define

$$H_k^t = \begin{bmatrix} u(0) & \cdots & u(t-k) \\ \vdots & & \vdots \\ u(k-1) & \cdots & u(t-1) \\ \hline y(0) & \cdots & y(t-k) \\ \vdots & & \vdots \\ y(k-1) & \cdots & y(t-1) \end{bmatrix}, \quad G_k^t = \begin{bmatrix} u(0) & \cdots & u(t-k) \\ \vdots & & \vdots \\ u(k-1) & \cdots & u(t-1) \\ \hline y(0) & \cdots & y(t-k) \\ \vdots & & \vdots \\ y(k-2) & \cdots & y(t-2) \end{bmatrix},$$

$$\ell_{\min}^t, \quad n_{\min}^t, \quad \text{and} \quad L_a^t := \min(L, N - n_{\min}^t + \ell_{\min}^t).$$

**Main idea:** start with  $k = 1$  and iterate between the following steps :

- increase the rank of  $G_k^t$  until no progress can be made
- increase the depth  $k$  by one

**Important question:** when to stop?

# Stopping criteria

**Simple observation:** We have that

$$\text{rank } \mathbf{G}_k^t \leq m + \text{rank } \mathbf{H}_{k-1}^t$$

**Lemma:** If

$$\text{rank } \mathbf{G}_k^t < m + \text{rank } \mathbf{H}_{k-1}^t,$$

then there exists an  $m - 1$  dimensional affine set  $\mathcal{A}^t \subseteq \mathbb{R}^m$  such that

$$\text{rank } \mathbf{G}_k^{t+1} = \text{rank } \mathbf{G}_k^t + 1 \quad \text{whenever} \quad u(t) \notin \mathcal{A}^t.$$

**Theorem:** Suppose that  $(u_{[0,t-1]}, y_{[0,t-1]})$  is such that

- $\mathbf{H}_k^t$  has full column rank, and
- $\text{rank } \mathbf{G}_k^t = m + \text{rank } \mathbf{H}_{k-1}^t$ .

Then,  $k = L_a^t + 1$  implies that

- 1  $k = L + 1$ ,
- 2  $t = T$ , and
- 3  $(u_{[0,T-1]}, y_{[0,T-1]})$  are **informative for Sysld**.

# The shortest experiment

## algorithm

```
1: procedure ONLINEEXPERIMENT( $L, N$ )
2:   choose  $u_{[0,m-1]}$  nonsingular
3:   measure outputs  $y_{[0,m-1]}$ 
4:    $t \leftarrow m, k \leftarrow 1$ 
5:   while  $k \neq L_a^t + 1$  do ▷ stopping criteria
6:      $k \leftarrow k + 1$ 
7:     if  $t = k - 1$  then ▷  $G_k^t$  has (full) rank 1
8:       choose  $u(t)$  arbitrarily
9:       measure output  $y(t)$ 
10:       $t \leftarrow t + 1$ 
11:    end if
12:    while rank  $G_k^t < m + \text{rank } H_{k-1}^t$  do ▷ rank  $G_k^{t+1} = \text{rank } G_k^t + 1$ 
13:      choose  $u(t) \notin \mathcal{A}^t$ 
14:      measure output  $y(t)$ 
15:       $t \leftarrow t + 1$ 
16:    end while
17:  end while
18:  return  $(u_{[0,t-1]}, y_{[0,t-1]})$  ▷  $(k, t) = (L + 1, T)$  and data are informative
19: end procedure
```

# The shortest experiment

## example

**True system and initial state:**

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Hence,  $n_{\text{true}} = 2$  and  $\ell_{\text{true}} = 1$ . We take  $N = L = 2$ .

$$u(0) = 1 \implies y(0) = \begin{bmatrix} -1 \\ 2 \end{bmatrix}. \quad \text{Let } t = 1 \text{ and } k = 1.$$

$$n_{\min}^1 = 0 \text{ and } \ell_{\min}^1 = 0 \implies L_a^1 = \min(L, N - n_{\min}^1 + \ell_{\min}^1) = 2 \implies k \neq L_a^1 + 1$$

$$\text{Set } k = 2. \text{ Since } t = k - 1, \text{ let } u(1) = 0 \text{ (arbitrary)} \implies y(1) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Now **increase rank:**

$$G_2^3 = \begin{bmatrix} 1 & 0 \\ 0 & u(2) \\ -1 & 0 \\ 2 & 0 \end{bmatrix}$$

# The shortest experiment

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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Now increase rank:

$$G_2^3 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \\ 2 & 0 \end{bmatrix}$$

# The shortest experiment

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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$$\text{Set } k = 2. \text{ Since } t = k - 1, \text{ let } u(1) = 0 \text{ (arbitrary)} \implies y(1) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Now increase rank:

$$G_2^4 = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & u(3) \\ -1 & 0 & 0 \\ 2 & 0 & 2 \end{bmatrix}$$



# The shortest experiment

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Hence,  $n_{\text{true}} = 2$  and  $\ell_{\text{true}} = 1$ . We take  $N = L = 2$ .

$$u(0) = 1 \implies y(0) = \begin{bmatrix} -1 \\ 2 \end{bmatrix}. \text{ Let } t = 1 \text{ and } k = 1.$$

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$$\text{Set } k = 2. \text{ Since } t = k - 1, \text{ let } u(1) = 0 \text{ (arbitrary)} \implies y(1) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Now increase rank:

$$G_2^4 = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \\ 2 & 0 & 2 \end{bmatrix}$$

# The shortest experiment

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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$$u(0) = 1 \implies y(0) = \begin{bmatrix} -1 \\ 2 \end{bmatrix}. \quad \text{Let } t = 1 \text{ and } k = 1.$$

$$n_{\min}^1 = 0 \text{ and } \ell_{\min}^1 = 0 \implies L_a^1 = \min(L, N - n_{\min}^1 + \ell_{\min}^1) = 2 \implies k \neq L_a^1 + 1$$

$$\text{Set } k = 2. \text{ Since } t = k - 1, \text{ let } u(1) = 0 \text{ (arbitrary)} \implies y(1) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Now increase rank:

$$G_2^5 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & u(4) \\ -1 & 0 & 0 & 0 \\ 2 & 0 & 2 & 1 \end{bmatrix}$$

# The shortest experiment

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

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$$u(0) = 1 \implies y(0) = \begin{bmatrix} -1 \\ 2 \end{bmatrix}. \quad \text{Let } t = 1 \text{ and } k = 1.$$

$$n_{\min}^1 = 0 \text{ and } \ell_{\min}^1 = 0 \implies L_a^1 = \min(L, N - n_{\min}^1 + \ell_{\min}^1) = 2 \implies k \neq L_a^1 + 1$$

$$\text{Set } k = 2. \text{ Since } t = k - 1, \text{ let } u(1) = 0 \text{ (arbitrary)} \implies y(1) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Now increase rank:

$$G_2^5 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ 2 & 0 & 2 & 1 \end{bmatrix}$$

# The shortest experiment

## example

True system and initial state:

$$A_{\text{true}} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad B_{\text{true}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad C_{\text{true}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D_{\text{true}} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}, \quad x(0) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Hence,  $n_{\text{true}} = 2$  and  $\ell_{\text{true}} = 1$ . We take  $N = L = 2$ .

$$\text{rank } H_1^5 = \text{rank} \begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 & 1 \\ 2 & 0 & 2 & 1 & 1 \end{bmatrix} = 3 \implies \text{rank } G_2^5 = 1 + \text{rank } H_1^5$$

$$\ell_{\min}^5 = 1 \text{ and } n_{\min}^5 = 2 \implies L_a^5 = \min(2, 2 - 2 + 1) = 1 \implies k = L_a^5 + 1.$$

**Conclusion:** The data  $(u_{[0,4]}, y_{[0,4]})$  are **informative for SysId**

**Reduction in # samples:** from  $T = 9$  to  $T = 7$  to  $T = 5$

# Conclusions

The shortest experiments for system identification require:

- 1 **Online design** of the inputs
  - 2 **Online adaptation** of the **depth** of the Hankel matrix
- 

Online design using depth- $(L + 1)$  Hankel matrix is shortest only if

$$L = N - n_{\text{true}} + \ell_{\text{true}}$$

---

**Final example:** For a system with

$$m = 80, \quad p = 10, \quad \ell_{\text{true}} = 20, \quad n_{\text{true}} = 100,$$

and

$$L = 100, \quad N = 150,$$

- **fundamental lemma** requires:  $T = 20330$
- **online design** (fixed depth) requires:  $T = 8280$
- **the shortest experiment** requires:  $T = 5850$

**Thank you!**

# Minimum lag and state dimension

$$\ell_{\min} = \min\{\ell \geq 0 \mid \exists n \geq 0 \text{ s.t. } \mathcal{E}(\ell, n) \neq \emptyset\}$$

minimum lag to explain the data

$$n_{\min} = \min\{n \geq 0 \mid \mathcal{E}(n) \neq \emptyset\}$$

minimum state dimension to explain the data

---

**Question:** How can we obtain  $\ell_{\min}$  and  $n_{\min}$  from the data?

---

Important role played by the **Hankel matrices**:

$$H_k = \begin{bmatrix} u(0) & \cdots & u(T-k) \\ \vdots & & \vdots \\ u(k-1) & \cdots & u(T-1) \\ \hline y(0) & \cdots & y(T-k) \\ \vdots & & \vdots \\ y(k-1) & \cdots & y(T-1) \end{bmatrix} \quad \text{and} \quad G_k = \begin{bmatrix} u(0) & \cdots & u(T-k) \\ \vdots & & \vdots \\ u(k-1) & \cdots & u(T-1) \\ \hline y(0) & \cdots & y(T-k) \\ \vdots & & \vdots \\ y(k-2) & \cdots & y(T-2) \end{bmatrix}$$

# Minimum lag and state dimension

$\ell_{\min} = \min\{\ell \geq 0 \mid \exists n \geq 0 \text{ s.t. } \mathcal{E}(\ell, n) \neq \emptyset\}$       minimum lag to explain the data

$n_{\min} = \min\{n \geq 0 \mid \mathcal{E}(n) \neq \emptyset\}$       minimum state dimension to explain the data

---

**Assumption:**       $u_{[0, T-1]} \neq 0_{m, T}$       (necessary for SysId)

---

We define for  $k \in [0, T-1]$ :

$$\delta_k = \text{rank } H_{k+1} - \text{rank } G_{k+1}$$

Then  $p \geq \delta_0 \geq \dots \geq \delta_{T-1} = 0$

---

$q$  := the smallest integer such that  $\delta_q$  is zero.       $q \in [0, T-1]$

---

**Theorem:**  $\ell_{\min} = q$  and  $n_{\min} = \sum_{i=0}^{\ell_{\min}} \delta_i$ .

---

**Theorem:**  $\mathcal{E}(\ell, n) \neq \emptyset \implies n - \ell \geq n_{\min} - \ell_{\min}$ .