

Data driven simulation with applications to system identification

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Notation

- discrete-time **LTI system** Σ

$$\sigma x = Ax + Bu, \quad y = Cx + Du,$$

where $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^m$, $y(t) \in \mathbb{R}^p$, and $(\sigma f)(t) = f(t+1)$

- \mathcal{B}_T — set of all T samples long trajectories $w := (u, y)$ of Σ ,

$$\mathcal{B}_T := \left\{ w := (u, y) \in (\mathbb{R}^m)^T \times (\mathbb{R}^p)^T \mid \right. \\ \left. \exists x \in (\mathbb{R}^n)^T \text{ s.t. } \sigma x = Ax + Bu, y = Cx + Du \right\}$$

- lag 1** of Σ — observability index of (A, C)

$n, m, p, 1, \mathcal{B}$ have the same meaning throughout



Outline

- 1 Introduction
- 2 Data driven simulation
- 3 Application for system identification



Problem formulation

Problem (Data driven simulation)

given:

- 1 trajectory $(\tilde{u}, \tilde{y}) \in \mathcal{B}_T$ of an LTI system Σ
- 2 upper bound n_{\max} of the order n
- 3 upper bound l_{\max} of the lag 1
- 4 time series $u_f \in (\mathbb{R}^m)^L$, where $L \in \mathbb{N}$

find: the response y_f of Σ under zero initial conditions and input u_f



Initial conditions

Setting zero initial conditions

Let

$$\left(\begin{bmatrix} u_p \\ u_f \end{bmatrix}, \begin{bmatrix} y_p \\ y_f \end{bmatrix} \right) \in \mathcal{B}_T$$

where $(u_p, y_p) = 0$ is at least 1 samples long zero sequence.

Then y_f is a zero initial conditions response due to u_f .



Fundamental Lemma

$$\begin{bmatrix} \tilde{u}(1) & \tilde{u}(2) & \cdots & \tilde{u}(T-\Delta+1) \\ \tilde{u}(2) & \tilde{u}(3) & \cdots & \tilde{u}(T-\Delta+2) \\ \vdots & \vdots & & \vdots \\ \tilde{u}(\Delta) & \tilde{u}(\Delta+1) & \cdots & \tilde{u}(T) \\ \hline \tilde{y}(1) & \tilde{y}(2) & \cdots & \tilde{y}(T-\Delta+1) \\ \tilde{y}(2) & \tilde{y}(3) & \cdots & \tilde{y}(T-\Delta+2) \\ \vdots & \vdots & & \vdots \\ \tilde{y}(\Delta) & \tilde{y}(\Delta+1) & \cdots & \tilde{y}(T) \end{bmatrix}$$

Every linear combination of the columns of $\mathcal{H}_\Delta(\tilde{w})$ is a response.

Under what conditions is every response generated that way?



Fundamental lemma

Let

$$\mathcal{H}_\Delta(\tilde{u}) = \begin{bmatrix} \tilde{u}(1) & \tilde{u}(2) & \cdots & \tilde{u}(T-\Delta+1) \\ \tilde{u}(2) & \tilde{u}(3) & \cdots & \tilde{u}(T-\Delta+2) \\ \vdots & \vdots & & \vdots \\ \tilde{u}(\Delta) & \tilde{u}(\Delta+1) & \cdots & \tilde{u}(T) \end{bmatrix}$$

\tilde{u} is **persistently exciting of order Δ** if $\mathcal{H}_\Delta(\tilde{u})$ is of full row rank.

Fundamental Lemma

Assume that

- 1 the LTI system Σ is controllable,
- 2 \tilde{u} is persistently exciting of order $\Delta + n$, and
- 3 (\tilde{u}, \tilde{y}) is a trajectory of Σ , i.e., $(\tilde{u}, \tilde{y}) \in \mathcal{B}_T$.

Then

$$\text{image} \left(\begin{bmatrix} \mathcal{H}_\Delta(\tilde{u}) \\ \mathcal{H}_\Delta(\tilde{y}) \end{bmatrix} \right) = \mathcal{B}_\Delta.$$

Algorithms for data driven simulation

Define: $\begin{bmatrix} \mathcal{H}_{n+t}(u) \\ \mathcal{H}_{n+t}(y) \end{bmatrix} =: \begin{bmatrix} u_p \\ u_f \\ y_p \\ y_f \end{bmatrix}$ $\begin{matrix} \text{row dim}(U_p) = \text{row dim}(Y_p) = n \\ \text{row dim}(U_f) = \text{row dim}(Y_f) = t \end{matrix}$

Theorem

Let Σ be controllable, $(\tilde{u}, \tilde{y}) \in \Sigma$, and \tilde{u} be persistently exciting of order $\Delta + 1_{\max} + n_{\max}$. Then the system of equations

$$\begin{bmatrix} U_p \\ U_f \\ Y_p \end{bmatrix} g = \begin{bmatrix} 0 \\ u_f \\ 0 \end{bmatrix}, \quad (1)$$

is solvable for any u_f and

$$y_f = Y_f \bar{g}$$

generates any zero initial conditions response.

Block algorithm

Algorithm

Input: \tilde{u} , \tilde{y} , n_{\max} , l_{\max} , and u_f .

- 1: Solve the system of equations (1) and let \bar{g} be a solution.
- 2: Compute $y_f = Y_f \bar{g}$.

Output: the response y_f of Σ to zero initial conditions and input u_f .

Limitation

$$t \leq \frac{T+1}{m+1} - l_{\max} - n_{\max}.$$

This can be avoided by “weaving” responses.



Iterative algorithm

Algorithm

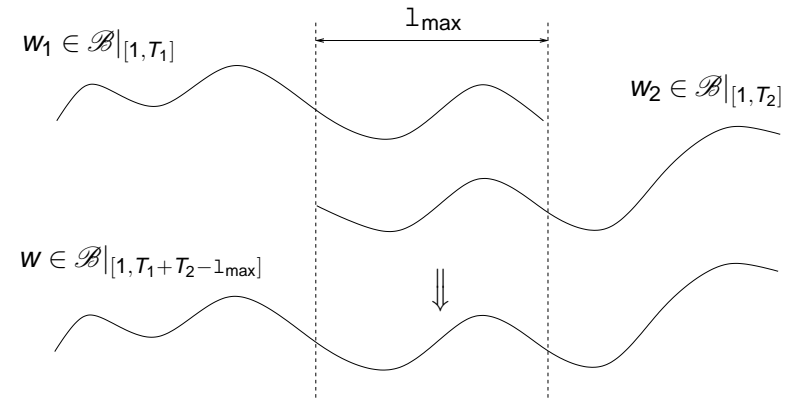
Input: \tilde{u} , \tilde{y} , n_{\max} , l_{\max} , u_f , and Δ satisfying the conditions of Theorem 2.

- 1: Set $k := 0$, $f_u^{(0)} := \begin{bmatrix} 0_{1 \times m \times 1} \\ u_f(1:\Delta) \end{bmatrix}$ and $f_{y,p}^{(0)} := 0_{1 \times m \times p \times 1}$.
- 2: **repeat**
- 3: Solve $\begin{bmatrix} U_p \\ U_f \\ Y_p \end{bmatrix} g^{(k)} = \begin{bmatrix} f_u^{(k)} \\ f_{y,p}^{(k)} \end{bmatrix}$ and let $\bar{g}^{(k)}$ be the solution found.
- 4: Compute the response $y_f^{(k)} := Y_f \bar{g}^{(k)}$.
- 5: $f_u^{(k+1)} := \begin{bmatrix} \sigma^\Delta f_u^{(k)} \\ u(k\Delta+1:(k+1)\Delta) \end{bmatrix}$, $f_{y,p}^{(k+1)} := \sigma^\Delta \begin{bmatrix} f_{y,p}^{(k)} \\ y_f^{(k)} \end{bmatrix}$.
- 6: $k := k + 1$
- 7: **until** $t < k\Delta$

Output: $y_f := \text{col}(y_f^{(0)}, \dots, y_f^{(k-1)})$.



Weaving responses



Simulation example $\tilde{w} \mapsto$ impulse response H

Simulation setup

- \mathcal{B} is of order $n = 4$, lag $l = 2$, with $m = 2$ inputs, and $p = 2$ outputs
- \tilde{w} is a trajectory of \mathcal{B} with length $T = 500$

Compared algorithms

- the block algorithm
- an iterative refinement of the block algorithm
- the function `impulse` from the Identification Toolbox of MATLAB

Approximation error $e = \|H - \hat{H}\|_F$ and execution time

method	error, e	time, sec.
block algorithm	10^{-14}	0.293
iterative algorithm	10^{-14}	0.066
impulse	0.059	0.584

Computation of free responses

The orthogonal projection, used in the MOESP algorithms,

$$Y_0 := \mathcal{H}_{n_{\max}}(\tilde{y})\Pi_{\tilde{u}}^{\perp}, \quad (2)$$

where

$$\Pi_{\tilde{u}}^{\perp} := I - \mathcal{H}_{n_{\max}}^{\top}(\tilde{u})(\mathcal{H}_{n_{\max}}(\tilde{u})\mathcal{H}_{n_{\max}}^{\top}(\tilde{u}))^{-1}\mathcal{H}_{n_{\max}}(\tilde{u}),$$

is a way to compute free responses from data. Observe that

$$\begin{bmatrix} \mathcal{H}_{n_{\max}}(\tilde{u}) \\ \mathcal{H}_{n_{\max}}(\tilde{y}) \end{bmatrix} \Pi_{\tilde{u}}^{\perp} = \begin{bmatrix} 0 \\ Y_0 \end{bmatrix}.$$



Conclusions

- system responses can be computed directly from (exact) data
- the algorithms require solving a linear system of equations
- the iterative algorithm weaves consecutive pieces of the response
- data driven simulation is relevant for system identification
- the orthogonal projection is a tool for computing free responses
- the oblique projection is a tool for computing sequential free responses



Computation of sequential free responses

The oblique projection, used in the N4SID algorithms,

$$Y_0 := Y_f / U_f W_p := Y_f \Pi_{\text{obl}}, \quad (3)$$

where

$$\Pi_{\text{obl}} := \begin{bmatrix} W_p^{\top} & U_f^{\top} \end{bmatrix} \begin{bmatrix} W_p W_p^{\top} & W_p U_f^{\top} \\ U_f W_p^{\top} & U_f U_f^{\top} \end{bmatrix}^+ \begin{bmatrix} W_p \\ 0 \end{bmatrix},$$

is a way to compute free responses, which initial conditions form a state sequence. Observe that

$$\begin{bmatrix} W_p \\ U_f \\ Y_f \end{bmatrix} \Pi_{\text{obl}} = \begin{bmatrix} W_p \\ 0 \\ Y_0 \end{bmatrix}.$$

