

ELEC system identification workshop

Exercises

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Outline

Introduction

Behavioral approach

Subspace methods

Optimization methods

SLRA package

Line fitting

problem: fit points $d_1, \dots, d_N \in \mathbb{R}^2$ by a line

1. find condition for existence of a line (any line in \mathbb{R}^2) that passes through the points
2. how would you test the condition in MATLAB?
3. implement a method for exact line fitting

Conic section fitting

problem: fit points $d_1, \dots, d_N \in \mathbb{R}^2$ by conic section

$$\mathcal{B}(S, u, v) = \{d \in \mathbb{R}^2 \mid d^\top S d + u^\top d + v = 0\}$$

1. find condition for existence of an exact fit
2. propose numerical method for exact fitting
3. implement the method and test it on the data

$$d_1 = \begin{bmatrix} -1 \\ -1 \end{bmatrix}, \quad d_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad d_3 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad d_4 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

Recursive sequence fitting

problem: fit $w = (w(1), \dots, w(T))$ by model

$$\mathcal{B} = \{ w \mid R_0 w + R_1 \sigma w + \dots + R_\ell \sigma^\ell w = 0 \}$$

1. find condition for existence of an exact fit
first, with, and then, without knowledge of ℓ
2. propose numerical method for exact fitting
find the smallest ℓ , for which exact model exists
3. implement the method and test it on the data

(1, 2, 4, 7, 13, 24, 44, 81)

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Checking whether a sequence is trajectory

1. given sequence w and polynomial R , propose method for checking numerically whether $w \in \mathcal{B} = \ker(R(\sigma))$
2. implement it in a function `w_in_ker(w, r)`
3. test it on the trajectory

$$w = (u_d, y_d) = ((0, 1), (0, 1), (0, 1), (0, 1))$$

and the system

$$\mathcal{B} = \ker(R(\sigma)), \quad R(z) = \begin{bmatrix} 1 & -1 \end{bmatrix} + \begin{bmatrix} -1 & 1 \end{bmatrix} z$$

Transfer function \mapsto kernel representation

1. what model $\mathcal{B}_{\text{tf}}(H)$ is specified by transfer function

$$H(z) = \frac{q(z)}{p(z)} = \frac{q_0 + q_1 z^1 + \cdots + q_\ell z^\ell}{p_0 + p_1 z^1 + \cdots + p_\ell z^\ell}$$

2. find R , such that

$$\mathcal{B}_{\text{tf}}(H) = \ker(R)$$

3. write function tf2r converting H (tf object) to R
and function r2tf converting R to H

Initial conditions specification by trajectory

LSIM(SYS, U, T, X0) specifies the initial state vector X0 at time T(1)
(**for** state-space models only).

problem: given minimal $\mathcal{B} = \mathcal{B}(A, B, C, D) \in \mathcal{L}_{m,l}$

1. show that $\underbrace{(w(-l+1), \dots, w(0))}_{w_p} \in \mathcal{B}$ determines $x(0)$
2. explain how to use w_p to "set" given $x(0)$
3. implement and test $w_p \leftrightarrow x(0)$ ($w_p \leftrightarrow x(0)$ / $x(0) \leftrightarrow w_p$)

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Exact identification of a kernel representation

let $w \in \mathcal{B} \in \mathcal{L}_{1,l}^2$ (SISO system)

implement the method $w \mapsto R$ (slide 19)

test it on examples (use `drss`)

Impulse response estimation

let $w \in \mathcal{B} \in \mathcal{L}_{1,l}^2$ (SISO system)

implement the method $w \mapsto H$ (slide 20–21)

test it on examples (use `drss`)

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Misfit computation using image repr.

given

- ▶ data $w = (w(1), \dots, w(T))$ and
- ▶ LTI system $\mathcal{B} = \text{image}(P(\sigma))$

derive method for computing

$$\text{misfit}(w, \mathcal{B}) := \min_{\hat{w} \in \mathcal{B}} \|w - \hat{w}\|_2$$

i.e., find the orthogonal projection of w on \mathcal{B}

Misfit computation using I/S/O representation

given

- ▶ data $w = (w(1), \dots, w(T))$ and
- ▶ LTI system $\mathcal{B} = \mathcal{B}(A, B, C, D)$

derive method for computing

$$\text{misfit}(w, \mathcal{B}) := \min_{\hat{w} \in \mathcal{B}} \|w - \hat{w}\|_2$$

i.e., find the orthogonal projection of w on \mathcal{B}

Latency computation using kernel repr.

given

- ▶ data w and
- ▶ LTI system $\mathcal{B}_{\text{ext}} = \ker(R(\sigma))$ $(w_{\text{ext}} := \begin{bmatrix} \hat{e} \\ w \end{bmatrix})$

find an algorithm for computing

minimize over e $\|\hat{e}\|$ subject to $(\hat{e}, w) \in \mathcal{B}_{\text{ext}}$

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Software

mosaic-Hankel low-rank approximation

<http://slra.github.io/software.html>

```
[sysh, info, wh] = ident(w, m, ell, opt)
```

- ▶ `sysh` — I/S/O representation of the identified model
- ▶ `opt.sys0` — I/S/O repr. of initial approximation
- ▶ `opt.wini` — initial conditions
- ▶ `opt.exct` — exact variables
- ▶ `info.Rh` — parameter R of kernel repr.
- ▶ `info.M` — misfit

```
[M, wh, xini] = misfit(w, sysh, opt)
```

demo file

Variable permutation

verify that permutation of the variables doesn't change the optimal misfit

```
T = 100; n = 2; B0 = drss(n);  
u = randn(T, 1); y = lsim(B0, u) + 0.001 * randn(T, 1);  
[B1, info1] = ident([u y], 1, n); disp(info1.MSE)  
    2.9736e-05  
[B2, info2] = ident([y u], 1, n); disp(info2.MSE)  
    2.9736e-05  
disp(norm(B1 - inv(B2)))  
    5.8438e-12
```

Output error identification

verify that the results of `oe` and `ident` coincide

```
T = 100; n = 2; B0 = drss(n);  
u = randn(T, 1); y = lsim(B0, u) + 0.001 * randn(T, 1);  
opt = oeOptions('InitialCondition', 'estimate');  
B1 = oe(iddata(y, u), [n + 1 n 0], opt);  
B2 = ident([u y], 1, n, struct('exct', 1));  
norm(B1 - B2) / norm(B1)
```

ans =

```
1.4760e-07
```