



KTH Electrical Engineering

On Optimal Input Design in System Identification for Control

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Outline

1. Theory
2. MPC Example
3. Conclusions
4. Future Work



Introduction

Framework for experiment design in system identification for control.

- Objective:
Find a minimum variance input signal to be used in system identification experiment.
- Such that:
The control application specification is guaranteed when using the estimated model in the control design.



Notation

The model structure is parametrized by θ .

- True system is given by θ_0 .
- Estimated model is given by $\hat{\theta}$.



Application Set

- Application cost: $V_{app}(\theta)$ such that

$$V_{app}(\theta_0) = 0, \quad V'_{app}(\theta_0) = 0 \text{ and } V''_{app}(\theta_0) \succeq 0.$$

- Application specification

$$V_{app}(\theta) \leq \frac{1}{2\gamma}, \quad \gamma > 0.$$



Application Set (cont.)

- Acceptable parameter set

$$\mathcal{E}_{app} = \{\theta \mid V_{app}(\theta) \leq \frac{1}{2\gamma}\}.$$

- Ellipsoidal approximation

$$\mathcal{E}_{app} \approx \{\theta \mid (\theta - \theta_0)^T V''_{app}(\theta_0)(\theta - \theta_0) \leq \frac{1}{\gamma}\}.$$

Application Set (cont.)

Example 1

- System

$$y(t) = b_1^0 u(t-1) + b_2^0 u(t-2) + d(t), \quad |b_2^0/b_1^0| < 1.$$

- P-controller used to reject constant disturbance $d(t)$.

- ▶ Nominal closed loop poles

$$-\beta \text{ and } -\beta b_2/(\beta b_1 - b_2).$$

- ▶ Controller gain

$$K = \frac{\beta^2}{\beta b_1 - b_2}, \quad 0 < \beta \leq 1.$$

Application Set (cont.)

Example 1 (cont.)

- $V_{app}(\theta)$ is error in static gain of sensitivity function.
- Hessian of application cost

$$V''_{app}(\theta_0) = C \begin{bmatrix} \beta^2 & -\beta \\ -\beta & 1 \end{bmatrix}.$$

- ▶ Eigenvalues: $C(\beta^2 + 1)$ and 0.
- ▶ Eigenvectors: $[-\beta, 1]^T$ and $[1, \beta]^T$.

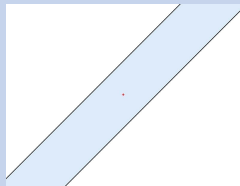


Figure: \mathcal{E}_{app} .



System Identification Set

Asymptotic quality property

$$\hat{\theta} \in \mathcal{E}_{SI} = \{\theta \mid (\theta - \theta_0)^T \mathbf{I}_F (\theta - \theta_0) \leq \eta\}.$$

(Key result from prediction error/maximum likelihood system identification.)



System Identification Set (cont.)

Example 1 (cont.)

The same system as before (first order FIR) yields

$$\mathbf{I}_F = \frac{1}{\lambda_e} \begin{bmatrix} r_0 & r_1 \\ r_1 & r_0 \end{bmatrix}, \quad r_\tau = E\{u(t)u(t-\tau)\}.$$

Optimal Input Signal Design

- Estimated parameters

$$\hat{\theta} \in \mathcal{E}_{SI} = \{\theta \mid (\theta - \theta_0)^T \mathbf{I}_F (\theta - \theta_0) \leq \eta\}.$$

- Acceptable parameters in application

$$\hat{\theta} \in \mathcal{E}_{app} \approx \{\theta \mid (\theta - \theta_0)^T V''_{app}(\theta_0) (\theta - \theta_0) \leq \frac{1}{\gamma}\}.$$

Minimize input power (r_0) subject to $\mathcal{E}_{SI} \subseteq \mathcal{E}_{app}$.
(Hjalmarsson, 2009)

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Minimize input power (r_0) subject to $\mathcal{E}_{SI} \subseteq \mathcal{E}_{app}$.
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Can be formulated as a convex problem!



Optimal Input Signal Design (cont.)

Optimization Problem

General problem formulation

$$\underset{r_\tau}{\text{minimize}} \ r_0, \text{ subject to } \mathcal{E}_{SI} \subseteq \mathcal{E}_{app}.$$

Optimal Input Signal Design (cont.)

Optimization Problem

General problem formulation

$$\underset{r_\tau}{\text{minimize}} \ r_0, \text{ subject to } \mathcal{E}_{SI} \subseteq \mathcal{E}_{app}.$$

Convex Optimization Problem (SDP/LMI)

Approximative problem formulation

$$\underset{r_\tau}{\text{minimize}} \ r_0, \text{ subject to } \mathbf{I}_F \succeq \eta \gamma V''_{app}(\theta_0).$$



Optimal Input Signal Design (cont.)

Geometric interpretation

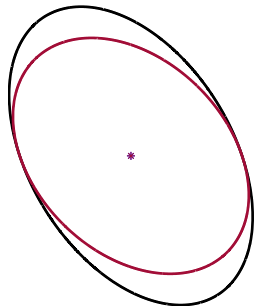
- minimize $r_0 \Leftrightarrow$ maximize \mathcal{E}_{SI} .
- $\mathbf{I}_F \succeq \eta\gamma V''_{app}(\theta_0) \Leftrightarrow \mathcal{E}_{SI} \subseteq \mathcal{E}_{app}$.

Optimal Input Signal Design (cont.)

Geometric interpretation

- minimize $r_0 \Leftrightarrow$ maximize \mathcal{E}_{SI} .
- $\mathbf{I}_F \succeq \eta \gamma V''_{app}(\theta_0) \Leftrightarrow \mathcal{E}_{SI} \subseteq \mathcal{E}_{app}$.

Figure: \mathcal{E}_{SI} (red) and \mathcal{E}_{app} (black). \rightarrow





Optimal Input Signal Design (cont.)

Example 1 (cont.)

- Optimization problem solved analytically, since only two parameters.
- Optimal input signal realized by an AR-process

$$u(t) = -\beta u(t-1) + e_u(t).$$

- Experiment design relates to application of model.
 - ▶ AR-pole $(-\beta) \Leftrightarrow$ nominal closed loop pole $(-\beta)$.

MPC Example

Example 2a

- System

$$y(t) = b_1^0 u(t-1) + b_2^0 u(t-2) + d(t).$$

- MPC used to reject constant disturbance $d(t)$.
 - ▶ MPC cost function

$$V_{MPC}(u(t)) = \sum_{t=1}^{T-1} y^2(t).$$

- ▶ MPC prediction horizon

$$T = 10.$$

MPC Example (cont.)

Example 2a (cont.)

- $V_{app}(\theta)$ is chosen as error in output signal.

- ▶ Application cost

$$V_{app}(\theta) = \frac{1}{M} \sum_{t=1}^M [y(t, \theta) - y(t, \theta_0)]^2.$$

- ▶ Number of measurements

$$M = 10.$$

- The Hessian of $V_{app}(\theta)$ is constructed numerically.

MPC Example (cont.)

Example 2a (cont.)

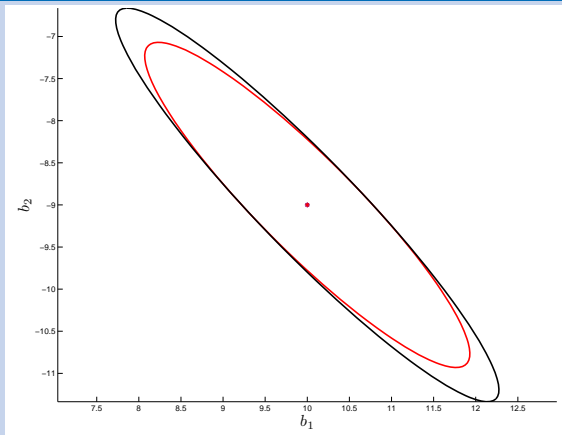


Figure: \mathcal{E}_{SI} (red) and \mathcal{E}_{app} (black).



MPC Example (cont.)

Example 2b

A cost is added on the input signal in the MPC.

- MPC cost function

$$V_{MPC}(u(t)) = \sum_{t=1}^{T-1} [y^2(t) + u^2(t)].$$

MPC Example (cont.)

Example 2b (cont.)

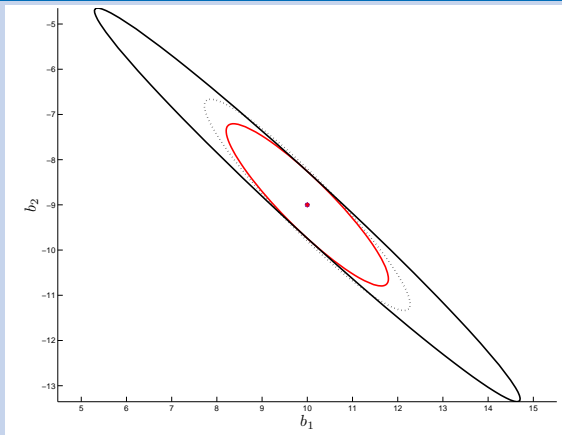


Figure: \mathcal{E}_{SI} (red), \mathcal{E}_{app} (black) and \mathcal{E}_{app} from 2a (dashed).



Conclusions

- Obtained estimated model guarantees (with high probability) that application requirements are met.
- Simple examples give insight in how experiment design relates to intended application of model.



Future Work

- How to choose V_{app} .
- Realistic MPC applications.
- Toolbox for optimal input design.