

On Optimal Input Design in System Identification for Model Predictive Control

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We present a method of performing optimal input design on a process controlled by MPC. Given a model structure and a measure of control performance degradation, the method provides an optimal input signal to be used in the identification experiment.

Optimal Input Design

Model:

$$\mathcal{M}(\theta) : \begin{aligned} x_{t+1} &= F(\theta)x_t + G(\theta)u_t + v_t, \\ y_t &= H(\theta)x_t + e_t, \\ v_t &\in \mathcal{N}(0, R_v), e_t \in \mathcal{N}(0, R_e) \end{aligned}$$

where $\theta \in \mathbb{R}^d$ is an unknown parameter vector. The true system is $\mathcal{S} = \mathcal{M}(\theta_0)$ for some $\theta_0 \in \mathbb{R}^d$.

Application: The relationship between θ and control performance degradation is given by $V_{app}(\theta)$. The allowed degradation is

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

yielding the approximative application set

$$\mathcal{E}_{app} = \{\theta \mid (\theta - \theta_0)^T V_{app}''(\theta_0) (\theta - \theta_0) \leq 1/\gamma\}.$$

Identification: The estimates $\hat{\theta}_N$ are obtained using PEM based on N observations. The confidence ellipsoid is

$$\hat{\theta}_N \in \mathcal{E}_{SI} = \{\theta \mid [\theta - \theta_0]^T \mathbf{P}^{-1} [\theta - \theta_0] \leq \chi_{\alpha}^2(d)/N\}, \text{ w. p. } \alpha,$$

where \mathbf{P} is the Fisher information matrix.

Optimization problem:

$$\begin{aligned} &\text{minimize}_{\phi_{u(\omega)}} \quad \text{experiment cost} \\ &\text{subject to} \quad \mathcal{E}_{SI} \subseteq \mathcal{E}_{app} \\ &\quad \quad \quad \phi_{u(\omega)} \geq 0 \quad \forall \omega \end{aligned}$$

where $\phi_{u(\omega)}$ is the input spectrum. An input is realized according to $\phi_{u(\omega)}$ and used in the identification experiment. The obtained $\hat{\theta}_N$ lie inside \mathcal{E}_{app} with, at least, probability α .

Experimental Identification Algorithm

Difficulties:

- θ_0 is unknown.
- $V_{app}(\theta)$ requires application based on models with more or less arbitrary θ .

Solutions:

- Use estimates instead of θ_0 in all the expressions.
- Evaluate $V_{app}(\theta)$ in simulation.

Identification algorithm:

- Step 0** Find an initial estimate of θ .
- Step 1** Evaluate $V_{app}(\theta)$ based on simulations of the model with the parameter estimates.
- Step 2** Design the optimal input signal based on \mathcal{E}_{app} and parameter estimates.
- Step 3** Find a new estimate of θ using the optimal input signal in the identification experiment.

Control Design for Water Process

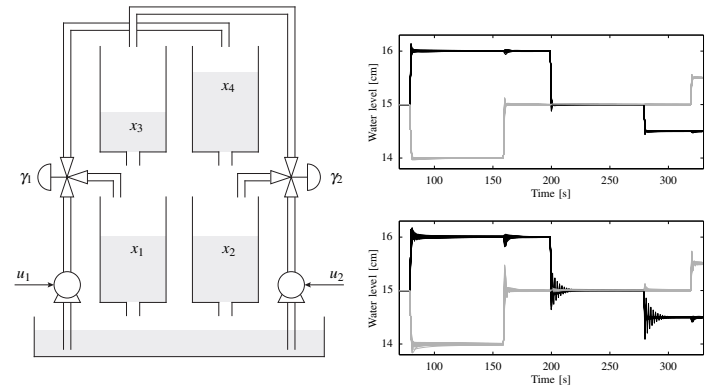


Figure 1: Water process. Figure 2: Trajectories. MPC controls the pump voltages, u_1 and u_2 . Lower tank levels, x_1 and x_2 , are outputs. Optimal and white input are used in upper and lower plot, respectively.

Application: The control objective is reference tracking of the lower tank levels using MPC. Thus, we choose

$$V_{app}(\theta) = \frac{1}{N} \sum_{t=1}^N \|y_t(\theta_0) - y_t(\theta)\|.$$

Experiment cost: Input power, trace $\left(\frac{1}{2\pi} \int_{-\pi}^{\pi} \phi_u(\omega) d\omega\right)$.

Result: 91 % of the trajectories have acceptable degradation for optimal input while only 15 % have for white input.

Conclusion

- Increased control performance.
- Linear framework applicable on nonlinear systems.

References

- [1] C. A. Larsson, M. Annergren, H. Hjalmarsson, "On Optimal Input Design in System Identification for Model Predictive Control", 50th IEEE CDC, 2011, to appear.
- [2] H. Hjalmarsson, "System Identification of Complex and Structured Systems", European Journal of Control, 2009.