





Optimal input design for online identification : a coupled observer-MPC approach

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IFAC'08, Korea, July, 7-10 2008

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Parametric sensitivity model Observer for state-affine systems

- **3. Proposed control approach** Model predictive control (MPC)
 - Final control structure

4. Case study

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Optimal input design

✓ Problem of parameter estimation needs to be addressed

 \checkmark A natural question: how can the input sequence be chosen in such a way that the parameters are optimally estimated?

✓ 1974 & 1985: studies on optimal input in linear systems

✓ 2002: Kessman and Stigter include analytical solution of optimal input design

✓ 2004: Stigter and Kessman have found recursive algorithm solutions

✓ But few studies exists for complex model based systems: nonlinear
 PDE system

Control problem tackled here: optimally control online a process based on a nonlinear PDE model to estimate online a set of the model parameters, under input/output constraints

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> Parametric sensitivity model

Let the model structure:

$$\begin{cases} \dot{x}(t) = f(x(t), \theta) + g(u(t), \theta) \\ y(t) = h(x(t)) \end{cases}$$
(1)

- x(t) : n-dimensional state vector
- u(t) : r-dimensional input vector
- y(t) : m-dimensional output vector
- θ : p-dimensional vector of time-invariant parameters

The associated parametric sensitivity model is derived from the model (1) :

$$\begin{cases} \dot{x}_{\theta}(t) \approx \frac{\partial f}{\partial x}(x(t),\theta).x_{\theta}(t) + \frac{\partial f}{\partial \theta}(x(t),\theta) + \frac{\partial g}{\partial \theta}(u(t),\theta) \\ y_{\theta}(t) \approx \frac{\partial h}{\partial x}(x(t)).x_{\theta}(t) \end{cases}$$

$$(2)$$

$$x_{\theta} \equiv \frac{\partial x}{\partial \theta} \quad \text{and} \quad y_{\theta} \equiv \frac{\partial y}{\partial \theta}$$

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Observer for state-affine systems [Hammouri and De Leon, 1990]

Observer : dynamic system obtained from the nominal model by adding a correction term.

Advantage: estimate the states not measured and unknown parameters.

The state affine system is in the following form :

$$\begin{cases} \dot{x} = A(u)x + B(u) \\ y = C(t)x \end{cases}$$
(3)

 $A(u) \in \mathfrak{R}^n \times \mathfrak{R}^n, B(u) \in \mathfrak{R}^r \times \mathfrak{R}^r etC(t) \in \mathfrak{R}^m \times \mathfrak{R}^n$

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Observer for state-affine systems [Hammouri and De Leon, 1990]

The observer for system (3) :

$$\begin{cases} \dot{\hat{x}} = A(u)\hat{x} + b(u) - S^{-1}C^{T}(C\hat{x} - y) \\ \dot{S} = -\rho S - A^{T}(u)S - SA(u) - C^{T}C \end{cases}$$
(4)

 $\rho \succ 0; \hat{x}(0) \in \Re^n$ is the initial condition and $S(t) \in PS(n)$,

Main idea : find a control law that maximizes y_{θ} , based on the model (1) coupled with the sensitivity function (2) and the observer (4).

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> Model Predictive Control (MPC)



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> Model Predictive Control (MPC)

Advantages:

✓ constraints (such as manipulated variables physical limitations, constraints due to operating procedures safety reasons...) may be specified

 \checkmark a model aims to predict the future behavior of the process and the best one is chosen by a correct optimal control of the manipulated variables

Drawbacks:

- ✓ computational time needed may limit online time
- \checkmark suboptimal solutions
- \checkmark how to handled unfeasibilities

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> Model Predictive Control (MPC)

$$\min_{u} J(u) = \sum_{j=k+1}^{j=k+p} \left[a(y_{ref}(j), y_p(j)) \right]$$

 $\begin{array}{l} \text{subject to constraints on the manipulated variable}:\\ u_{min} \leq u(j) \leq u_{max} \forall j \in \{\!\!\{k+1,k+p\}\!\!\}\\ \Delta u_{min} \leq u(j) - u(j-1) \leq \Delta u_{max} \forall j \in \{\!\!\{k+1,k+p\}\!\!\}\\ \text{subject to contraints on the controlled variable}:\\ c(\gamma_p(j),\gamma_{ref}(j)) \leq 0 \forall j \in \{\!\!\{k+1,k+p\}\!\!\} \end{array}$

The function **a** means: trajectory tracking, processing time minimization, energy consumption minimization, sensitivity

maximization, ...

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Model Predictive Control (MPC) [Dufour et al, IEEE TCST 11(5) 2003]

Originally developed for nonlinear PDE model control
 Main idea: decrease the online time needed to compute the PDE model based control

Approach:

- ✓ Input constraints: hyperbolic transformation
- ✓ Output constraints: exterior penalty method
- ✓ Linearization + sensitivities computed off line
- \checkmark One line use of a time varying linear model
- ✓ One line resolution of the penalized (and so unconstrained)
 optimization control problem: as modified Levenberg Marquart
 Algorithm
- ✓ The turnkey MPC@CB software is used

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Final control structure

To maximize the parametric sensitivity of the process output $\frac{\partial y_p}{\partial \theta}$

Cost function *J* to minimize:

$$J(u) = \sum_{j=k+1}^{j=k+N_p} \frac{1}{\left\| \mathcal{E} + \frac{\partial y_p}{\partial \hat{\theta}}(k) \right\|^2}$$

> Final control structure

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Linearized IMC-MPC observer based structure for coupled input design and parameter estimation



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> Model for powder coating curing process

Powder coatings = fine particles of: resin +cross-linker in thermosetting or thermoplastic powder coatings+ pigments + extenders + flow additives and fillers to achieve specific properties (color,...)



Schematic drawing of the "substrat + powder" sample

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> Model for powder coating curing process

- \checkmark Thermal model based on the Fourier law of heat conduction uses:
 - 1. the temperature variable varying in the thickness of the powder coating metal sample
 - 2. the degree of cure conversion variable (ranging from 0+ to 1 and the end)
- ✓ A non linear PDE Boundary control problem has to be tackled

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Model for powder coating curing process [Bombard et al, 2006]

$$\frac{\partial T_{p}(z,t)}{\partial t} = \frac{\lambda_{c,p}}{\rho_{p}C_{pp}} \frac{\partial^{2}T_{p}(z,t)}{\partial z^{2}} - \frac{e_{p}\Delta H_{0}}{C_{pp}} k_{0}e^{\left(\frac{-E_{a}}{RT_{p}(z,t)}\right)} x^{m}(1-x)^{n}$$

$$\forall z \in \mathbf{D}, e_{p} [\forall t > 0]$$

Tp(z,t) = temperature across the powder film thicknessep = film thickness(~0.1 mm)x(z,t) = degree of cure

$$\frac{\partial T_{s}(z,t)}{\partial t} = \frac{\lambda_{c,s}}{\rho_{s}C_{ps}} \frac{\partial^{2}T_{s}(z,t)}{\partial z^{2}} \qquad \forall z \in \mathbf{p}, e_{p} + e_{s} \left[\forall t > 0 \right]$$

Ts(z,†) = temperature across substrate es = film thickness(~ 1 mm)

> Model for powder coating curing process [Bombard et al, 2006] Outline trol problem 3 boundary conditions for the temperature: tatement eliminaries $-\lambda_{p} \frac{\partial T_{p}(z,t)}{\partial z} = \alpha_{p} \varphi_{ir}(t) - \sigma \varepsilon_{p} (T_{p}^{4}(z,t) - T_{ext}^{4}) - h_{p} (T_{p}(z,t) - T_{ext})$ Proposed at z = 0, $\forall t > 0$ rol approach Manipulated variable **Estimated** parameter ise study $-\lambda_{c,p}\frac{\partial T_p(z,t)}{\partial z} = -\lambda_{c,s}\frac{\partial T_p(z,t)}{\partial z} \quad \text{at } z = e_p, \forall t > 0$ PC@CB ware main eatures $-\lambda_{s}\frac{\partial T_{s}(z,t)}{\partial z} = -\sigma\varepsilon_{s}(T_{s}^{4}(z,t) - T_{ext}^{4}) - h_{s}(T_{s}(z,t) - T_{ext})$ mulation results at $z = e_p + e_s$, $\forall t > 0$ clusion L rspectives

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Model for powder coating curing process [Bombard et al, 2006]

The degree of cure x(z,t) of the powder:

$$\frac{\partial x(z,t)}{\partial t} = k_0 e^{\left(\frac{-E_a}{RT_p(z,t)}\right)} x^m (1-x)^n \qquad \forall z \in [0, e_p] \forall t > 0$$

Initial conditions:

$$T_p(z,t) = T_s(z,t) = T_{ext} \forall z \in [0, e_p + e_s], t = 0$$
$$x(z,t) = 0^+ \forall z \in [0, e_p], t = 0$$

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- ✓ Developed under Matlab, MPC@CB© solvers any users defined:
 - trajectory tracking problem
 - maximization of the parameter sensitivity
 - operating time minimization problem
 - any cost function
 - input/output constraint handled
- Any user defined continuous model (SISO, MISO, SIMO, MIMO model), including large scale PDE model
- \checkmark Easy to introduce a user defined observer
- \checkmark Easy to apply software for simulation or real time application

MPC@CB©: flexibility/ ease for a quick use in control!

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mulation results eclusion L rspectives Sensitivity of the process output, with magnitude+ velocity input constraints+ output constraint (sample time= 1s)





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Conclusions

 ✓ Approach coupling a process model, a parametric sensitivity model, an observer and a predictive controller was used on line for optimal constrained optimization

✓ Optimal identification of PDE system by a general MPC@CB© software has been shown

Perspectives

✓ Extended for other process models and for parameter vector case

✓ To use MPC@CB© : dufour@lagep.univ-lyon1.fr

Thank you

Any questions ?