Computer Science

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Systems Modelling and Analysis

Choose yourself and new technologies

L.13. Identification of dynamic plants



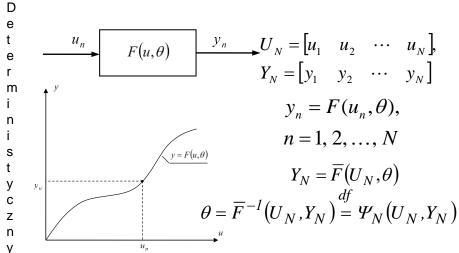




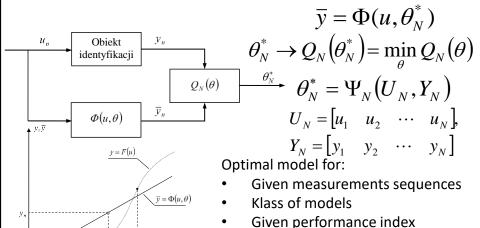
 $F(u,\theta,\omega)$

 $u_i = u_n$

Plant in the class of model

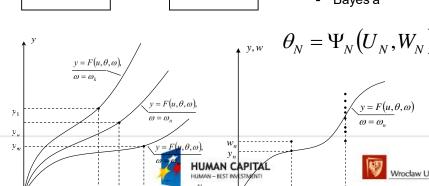


Choice of the best model

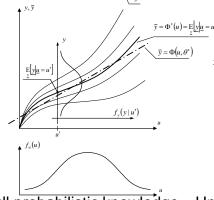




- Least square
 - Maximum
 - likelihood Bayes'a



h(y,z)

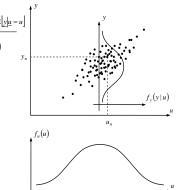


Full probabilistic knowledge

First type regrsesion

Project co-financed from the EU European Social Fund

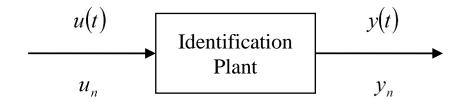
Second type regression



Unknown probabilostic knowledge

- Performance index estimatin
- arameter estimation of the probability distribution
- Probabilit ditribution estimatoin

Identification of Dynamic Plants



Descriptions:

- state variable;
- differential/difference equation;
- transfer functions: K(s), K(z);
- impulse response: $k_i(t)$, k_{in} ;
- step response: h(t), h_n

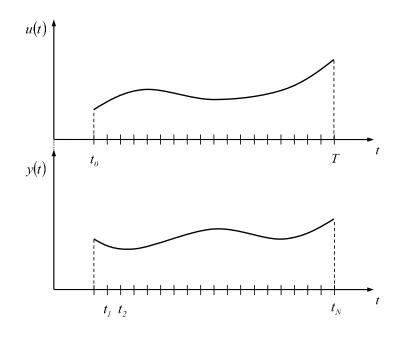






Identification of Dynamic Plants Continous Plant

For input signal $\{u(t)\}_{t_0}^T$ we measure respective output signal $\{y(t)\}_{t_0}^T$:



$$t_0 < t_1 < \ldots < t_N \le T$$

$$\{u(t_n)\}_{n=0}^N \qquad \{y(t_n)\}_{n=0}^N$$

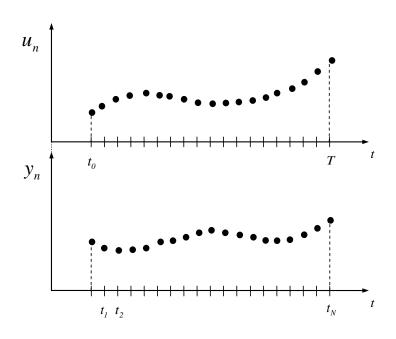






Identification of Dynamic Plants Discrete Plant

For input signal u_n we measure respective output signal y_n :



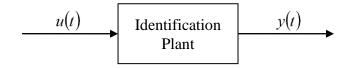
$$t_0 < t_1 < \ldots < t_N \le T$$

$$\{u_n\}_{n=0}^N \qquad \{y_n\}_{n=0}^N$$









Let us assume that plant characteristic is known:

$$F\left(\frac{d^{m}y(t)}{dt^{m}}, \frac{d^{m-1}y(t)}{dt^{m-1}}, \dots, \frac{dy(t)}{dt}, y(t); \frac{d^{v}u(t)}{dt^{v}}, \frac{d^{v-1}u(t)}{dt^{v-1}}, \dots, \frac{du(t)}{dt}, u(t); \theta\right) = 0, \qquad m \ge v$$

and initial conditions w_0 are given.

The problem is to determine parameters θ .







If u(t) is given **analytically**, e.g. $u(t) = \mathbf{1}(t)$, $u(t) = A\sin(\omega t)$; then it is possible to work out solution analytically:

$$y(t) = \mathscr{F}\left(\left\{u(\tau)\right\}_{\tau=t_0}^t, w_0; \theta\right)$$

$$y(t_n) = \mathscr{F}(\{u(\tau)\}_{\tau=t_0}^{t_n}, w_0; \theta), n = 1, 2, ..., N$$

Solution of the above system of equations with respect $\,\theta$ gives identification algorithm.







Example

$$\frac{dy(t)}{dt} = \theta u(t)$$

Input signal:

$$u(t) = \mathbf{1}(t)$$

Initial condition:

$$y(0) = 0$$

Solution:

$$y(t) = \theta \int_{0}^{t} u(\tau) d\tau = \theta \int_{0}^{t} 1(\tau) d\tau = \theta t$$

$$y(t_n) = \theta t_n$$
 $\theta = \frac{y(t_n)}{t}$

$$\theta = \frac{y(t_n)}{t}$$







Approximate numerical solution:

Measurements:

$$\{u(t_n)\}_{n=0}^N \qquad \{y(t_n)\}_{n=0}^N$$

Numerical solution:

$$\widetilde{y}(t_n, \theta) = \mathscr{F}_N(\{u(t_k)\}_{k=0}^n, w_0; \theta)$$
 $y(t_n) \approx \widetilde{y}(t_n, \theta)$

$$\theta \approx \theta_N \to \min \sum_{n=0}^{N} (y(t_n) - \tilde{y}(t_n, \theta))^2 = \min \sum_{n=0}^{N} (y(t_n) - \mathscr{T}_N(\{u(t_k)\}_{k=0}^n, w_0; \theta))^2$$

where:

 $B(\theta)$ – error of numerical procedure







Calculations:

$$\theta_{0} \to \widetilde{y}(t_{n}, \theta_{0}) \to B(\theta_{0})$$

$$\theta_{l} \to \widetilde{y}(t_{n}, \theta_{l}) \to B(\theta_{l})$$

$$\vdots$$

$$\theta_{K} \to \widetilde{y}(t_{n}, \theta_{K}) \to B(\theta_{K})$$

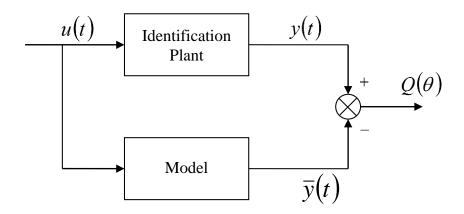
$$\theta_{K} \approx \theta$$

K – number of steps of the numerical optimization methods









$$\Phi\left(\frac{d^{m}\overline{y}(t)}{dt^{m}}, \frac{d^{m-l}\overline{y}(t)}{dt^{m-l}}, \dots, \frac{d\overline{y}(t)}{dt}, \overline{y}(t); \frac{d^{v}u(t)}{dt^{v}}, \frac{d^{v-l}u(t)}{dt^{v-l}}, \dots, \frac{du(t)}{dt}, u(t); \theta\right) = 0$$

 w_0 - initial conditions







1. u(t) is given **analytically**, e.g. $u(t) = \mathbf{1}(t)$, $u(t) = A\sin(\omega t)$;

Model:

$$\overline{y}(t,\theta) = \widetilde{\Phi}(\{u(\tau)\}_{\tau=t_0}^t, w_0; \theta)$$

Performance index:

$$Q(\theta) = \int_{t_0}^{T} (y(t) - \overline{y}(t, \theta))^2 dt = \int_{t_0}^{T} (y(t) - \widetilde{\Phi}(\{u(\tau)\}_{\tau=t_0}^t, w_0; \theta))^2 dt$$

Optimization problem:

$$\theta^* \to Q(\theta^*) = \min_{\theta \in \Theta} Q(\theta)$$







Example

Model:

$$\frac{d\overline{y}(t)}{dt} = \theta u(t)$$

Input signal: $u(t) = \mathbf{1}(t)$

Initial condition: y(0) = 0

Solution:

$$\overline{y}(t) = \theta \int_{0}^{t} u(\tau) d\tau = \theta t$$

Performence index: $Q(\theta) = \int_{0}^{T} (y(t) - \theta t)^{2} dt$

$$\theta^* = \frac{\int_0^T y(t)tdt}{\int_0^T t^2 dt}$$







2. Discrete measurements of output signal: $\{u(t)\}_{t_0}^T$, $\{y(t_n)\}_{n=0}^N$;

Performance index:

$$Q_N(\theta) = \sum_{n=0}^N \left(y(t_n) - \widetilde{\Phi}(\{u(\tau)\}_{\tau=t_0}^{t_n}, w_0; \theta) \right)^2$$

Optimization problem:

$$\theta_N^* \to Q_N(\theta_N^*) = \min_{\theta \in \Theta} Q_N(\theta)$$







Numerical method:

Measurements: $\{u(t_n)\}_{n=0}^N$, $\{y(t_n)\}_{n=0}^N$;

Solution:

$$\widetilde{\overline{y}}_{N}(t_{n},\theta) = \widetilde{\Phi}_{N}(\{u(t_{k})\}_{k=0}^{n}, w_{0}; \theta)$$

$$Q_{N}(\theta) = \sum_{n=0}^{N} (y(t_{n}) - \tilde{y}_{N}(t_{n}, \theta))^{2} = \sum_{n=0}^{N} (y(t_{n}) - \tilde{\Phi}_{N}(\{u(t_{k})\}_{k=0}^{n}, w_{0}; \theta))^{2}$$







We start from arbitrary taken θ_0 and then perform following calculations:

$$\theta_{0} \to \overline{y}(t_{n}, \theta_{0}) \to Q_{N}(\theta_{0})$$

$$\theta_{I} \to \overline{y}(t_{n}, \theta_{I}) \to Q_{N}(\theta_{I})$$

$$\vdots$$

$$\theta_{K} \to \overline{y}(t_{n}, \theta_{K}) \to Q_{N}(\theta_{K})$$

$$\theta_{K}^{*} \approx \theta_{N}^{*}$$







Thank you for attention

