Systems Dynamics

Course ID: 267MI - Fall 2018

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System Identification

Modelling Identification Prediction & filtering Disciplines providing tools to **estimate** variables and unknown parameters and to **design models** of natural and artificial systems using **experimental data**. 267MI -Fall 2018

Lecture 4
Model identification from data

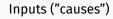
Why do we need models?

Constructing models for a slice of reality and studying their properties is really what science is about. The **models** – "the hypotheses", "the laws of nature", "the paradigms" – can be of a more or less formal character, but they all have the **fundamental property** that they try **to link observations to some pattern**.

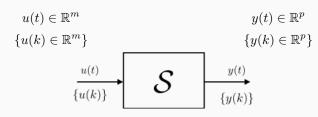
L. Ljung, T. Glad, "Modeling of Dynamic Systems", Prentice Hall, 1994

"Transparent box" modeling approach

So far, approach undertaken to devise dynamical models:



Outputs ("effects")



Definition of the "system" entity to be analysed

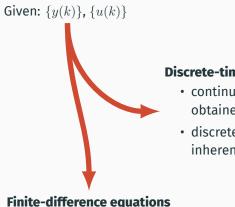
Physical laws, a priori knowledge, heuristic considerations. statistical evidence, etc.

Mathematical models: algebraic and/or differential and/or difference equations

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"Black-box" modeling approach (identification)



Discrete-time model:

- · continuous-time system and data obtained by sampling
- · discrete-time system and data inherently discrete-time

Finite-difference equations

A different approach (data-based)



Experimental data, sensor measurements, historical data, etc.

"Black-box" modeling approach

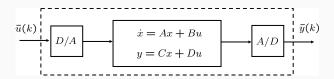
Input, output and disturbance variables are characterized in terms of numerical sequences. These are the data to be used to determine the dynamical model.

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Continuous-time system and data obtained by sampling

Linear, time-invariant case:



 $u(t) = \bar{u}(k)$

Recall the **step-invariant transform**

 $\bar{y}(k) = y(t_k)$

 $t_k \le t < t_{k+1}$

$$\begin{cases} \bar{x} \left[\left(k+1 \right) \right] = \bar{A}\bar{x} \left(k \right) + \bar{B}\bar{u} \left(k \right) \\ \bar{y} \left(k \right) = \bar{C}\bar{x} \left(k \right) + \bar{D}\bar{u} \left(k \right) \end{cases}$$

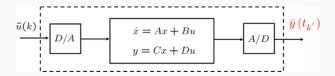
Letting $t_{k+1} - t_k = \Delta$

$$\bar{A} = e^{A\Delta}$$
 $\bar{B} = \int_0^\Delta e^{Ar} B \, dr$ $\bar{C} = C$

$$\bar{C} = C$$

$$\bar{D} = D$$

Continuous-time system and data obtained by sampling (cont.)



$$\begin{array}{ll} u(t) = \bar{u}(k) & \text{What if the output is sampled at} & \bar{y}(k) = y\left(t_{k'}\right) \\ t_k \leq t < t_{k+1} & t_{k'} \neq t_k \text{ with } t_k \leq t_{k'} < t_{k+1} \end{array} \textbf{?}$$

· Let's recall the expression

$$y(t) = C e^{A(t-t_0)} x(t_0) + C \int_{t_0}^t e^{A(t-\tau)} Bu(\tau) d\tau + Du(t)$$

for the movement of the output of a continuous-time LTI system (from "Fundamentals of Automatic Control").

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Continuous-time system and data obtained by sampling (cont.)

• If $t_{k'} \neq t_k$ then

$$\bar{C} = C e^{A\alpha}$$
 $\bar{D} = C \left(\int_0^\alpha e^{Ar} dr \right) B + D$ $t_{k'} - t_k = \alpha (< \Delta)$

• When $t_{k'}=t_k$ obviously

$$\bar{C} = C$$
 $\bar{D} = D$

• In both the cases the following expressions hold (remember the *step-invariant transform*)

$$\bar{A} = e^{A\Delta}$$
 $\bar{B} = \int_0^\Delta e^{Ar} B \, dr$ $\Delta = t_{k+1} - t_k \, \forall k$

Continuous-time system and data obtained by sampling (cont.)

• Let's consider t_k as initial time instant (i.e $t_0=t_k$), the instant $t_{k'}$ as final time instant and let's assume $t_{k'}-t_k=\alpha$. Recall also the stair-wise behavior of the input signal:

$$u(t) = \bar{u}(k), t_k \le t < t_{k+1}.$$

$$y(t_{k'}) = C e^{A\alpha} x(t_k) + C \left(\int_{t_k}^{t_{k'}} e^{A(t_{k'} - \tau)} B d\tau \right) u(t_k) + Du(t_k)$$

- Substitute $r=t_{k^{\prime}}-\tau\,$ into the integral term and rewrite the expression

$$y(t_{k'}) = C e^{A\alpha} x(t_k) + C \left(\int_0^\alpha e^{Ar} dr \right) Bu(t_k) + Du(t_k)$$

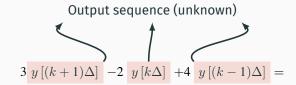
· Let's compare with the discrete-time output expression

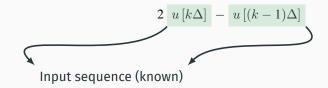
$$\bar{y}(k) = \bar{C}\bar{x}(k) + \bar{D}\bar{u}(k)$$

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"Black box" modeling: an example

As usual, let's assume Δ as the sampling time.





Typical notations

• With sampling-time Δ enhanced:

$$3y\left[(k+1)\Delta\right] - 2y\left[k\Delta\right] + 4y\left[(k-1)\Delta\right] = \\ 2u\left[k\Delta\right] - u\left[(k-1)\Delta\right]$$

· Compact:

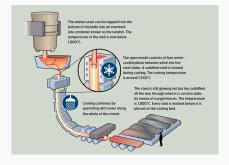
$$3y_{k+1} - 2y_k + 4y_{k-1} = 2u_k - u_{k-1}$$

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A real application: steel continuous casting





General expression

- Typical framework: linear finite-difference equations with constant coefficients.
- General expression takes on the form:

$$a_n y_{k+n} + a_{n-1} y_{k+n-1} + \dots + a_0 y_k =$$

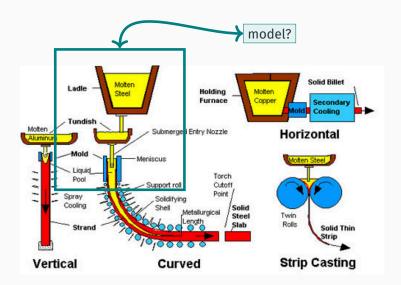
$$b_m u_{k+m} + b_{m-1} u_{k+m-1} + \dots + b_0 u_k$$

with given initial conditions

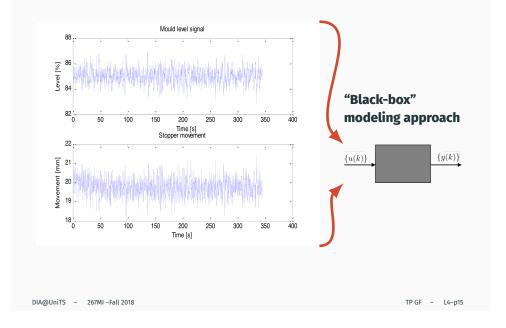
$$\{y_k: k=-n, -(n-1), \ldots, 0\}$$

and known input sequence u_k .

A real application: steel continuous casting (cont.)



A real application: steel continuous casting (cont.)



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