

# What are we doing in identification?

## 1 Example: the first order dynamic system

$$y(k) = 0.9512y(k-1) + 0.09754u(k-1)$$

1 *1st order*

1 *Regressors:  $y(k-1), u(k-1)$*

1  *$y(k) = -a_1y(k-1) + b_1u(k-1)$*

$$H(z) = \frac{0.09754z^{-1}}{1 - 0.9512z^{-1}}$$

$$\begin{bmatrix} y(2) \\ y(3) \\ \dots \end{bmatrix} = \begin{bmatrix} -y(1) & u(1) \\ -y(2) & u(2) \\ \dots & \dots \end{bmatrix} \begin{bmatrix} a_1 \\ b_1 \end{bmatrix} \quad \mathbf{y} = \Psi\theta$$

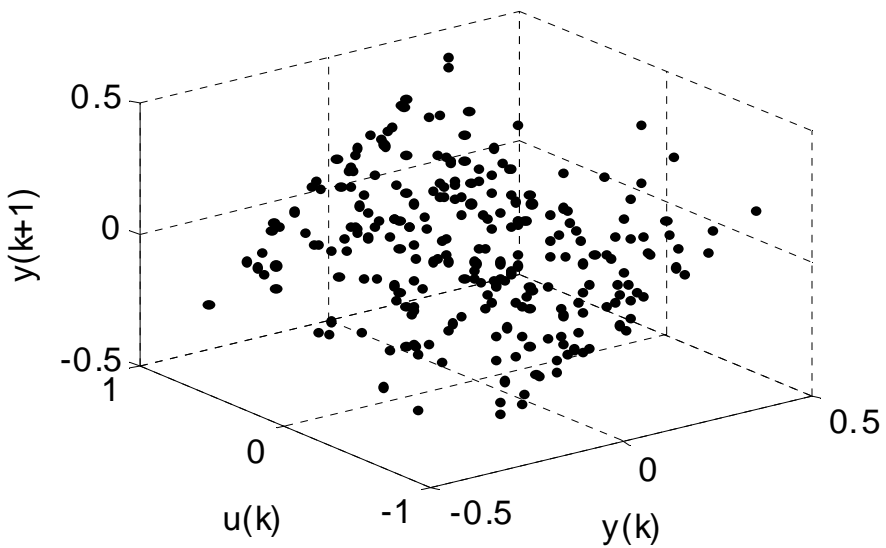
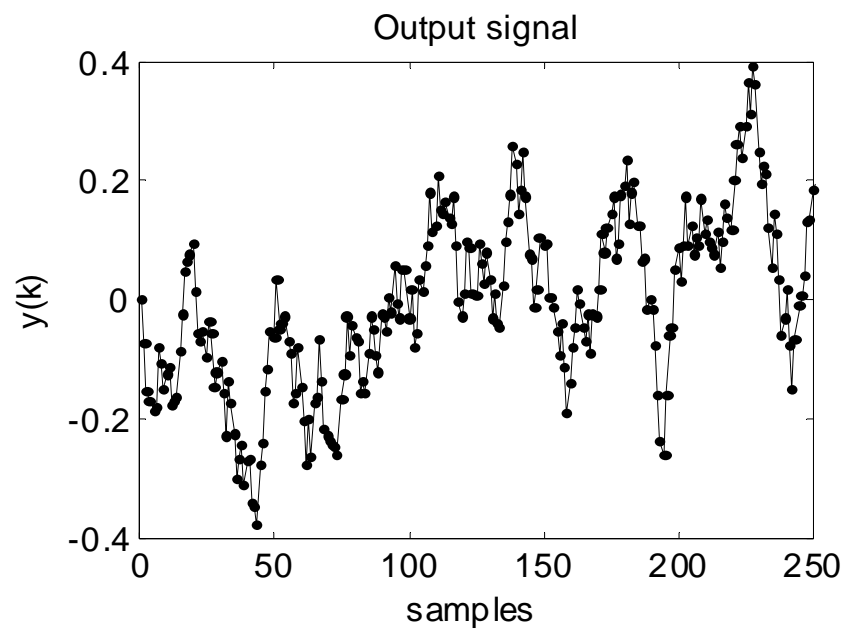
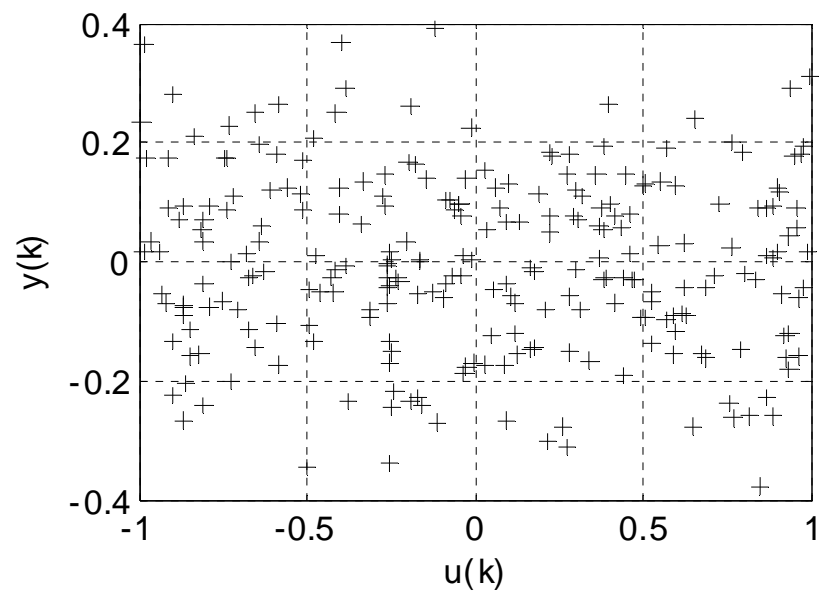
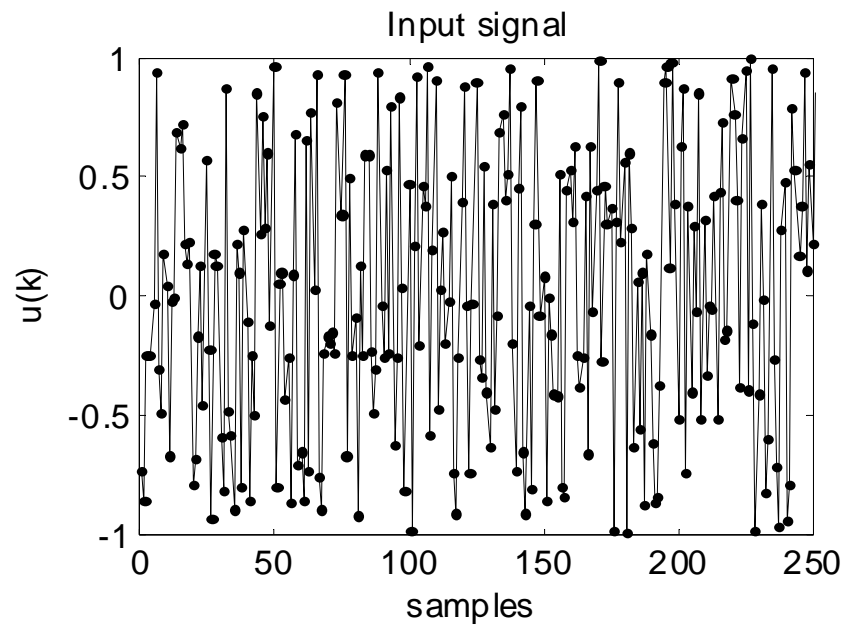
1 Order of rows and columns can be changed!!

1 Optimal solution by least squares cost function

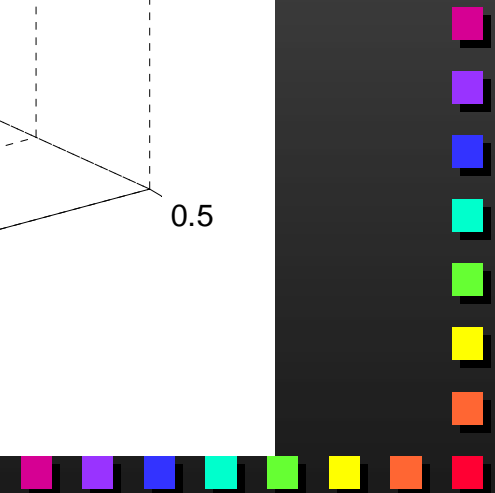
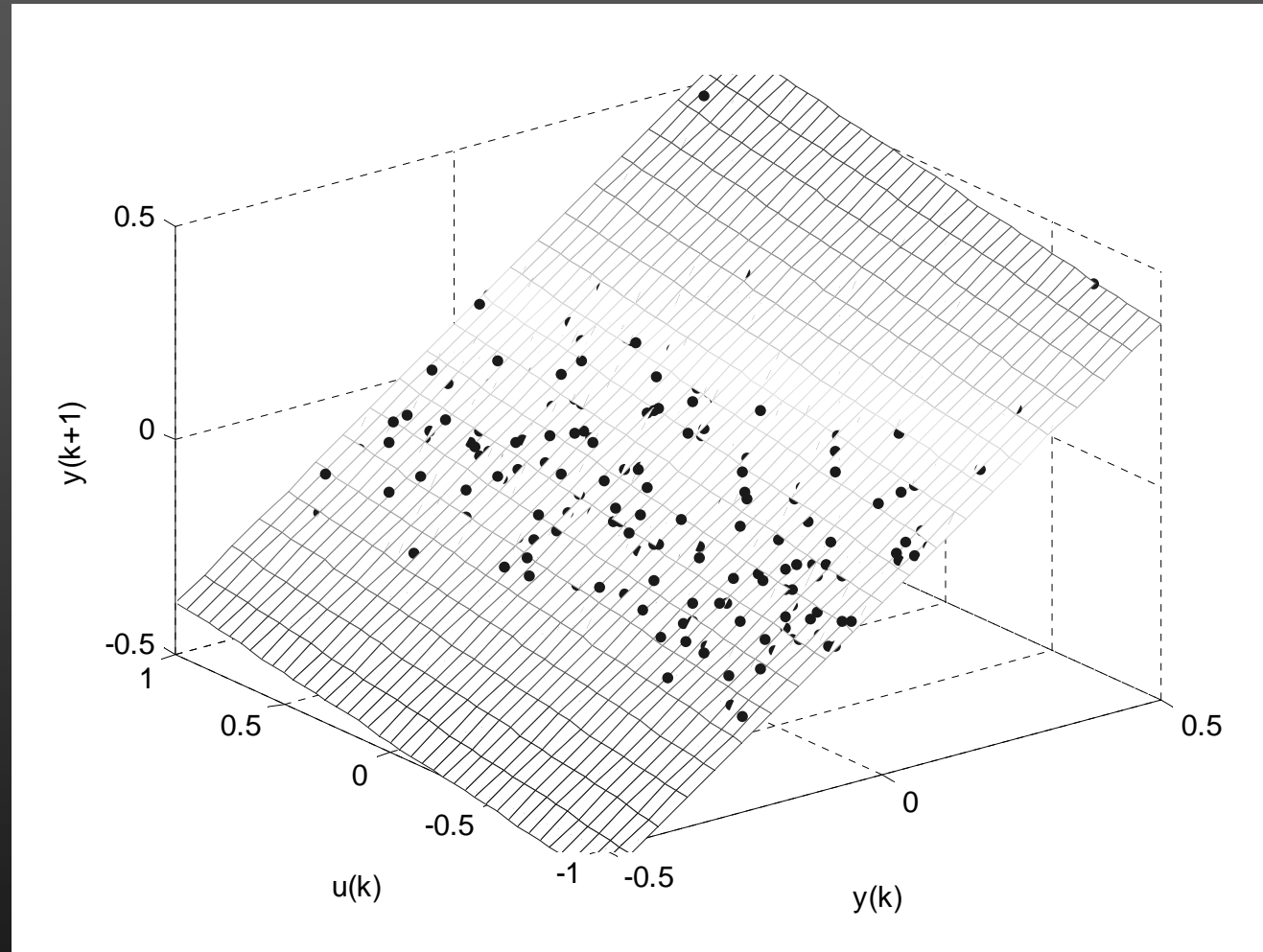
$$\underline{\theta} = (\Psi^T\Psi)^{-1}\Psi^T\mathbf{y}$$

1 Parameters are optimal for one-step-ahead prediction, validation is done with simulation (multi-step-ahead prediction).





# Model

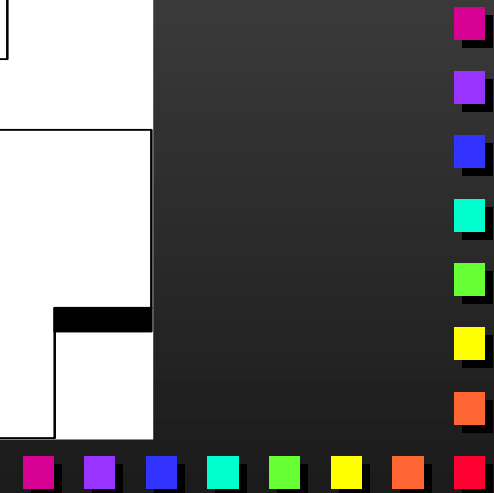
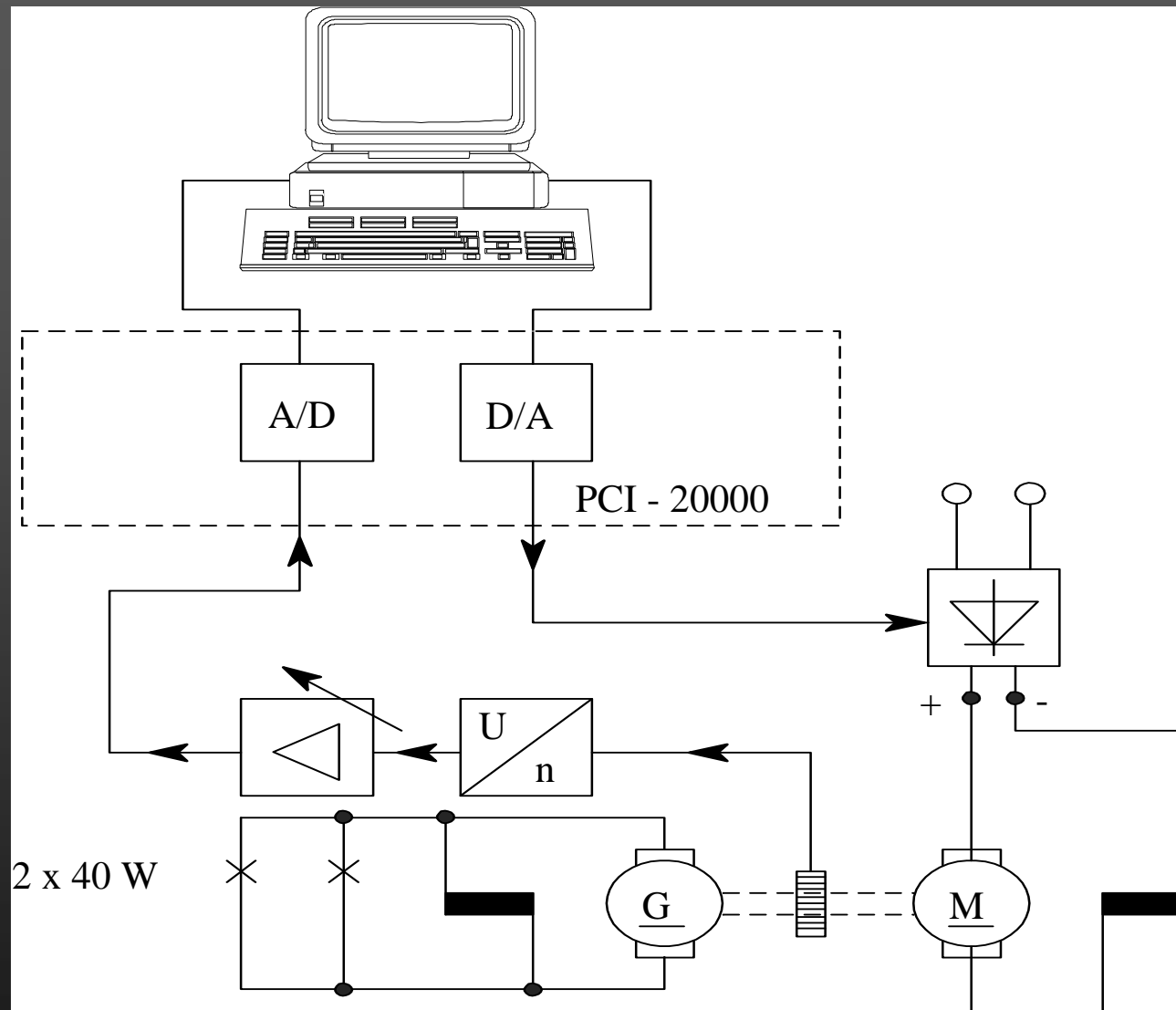




# System Identification Case Study



# Motor-generator test rig

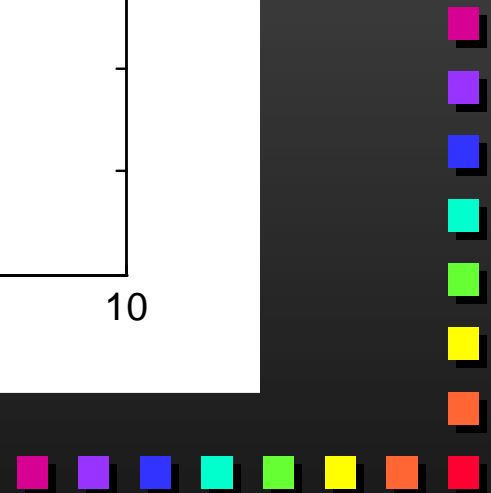
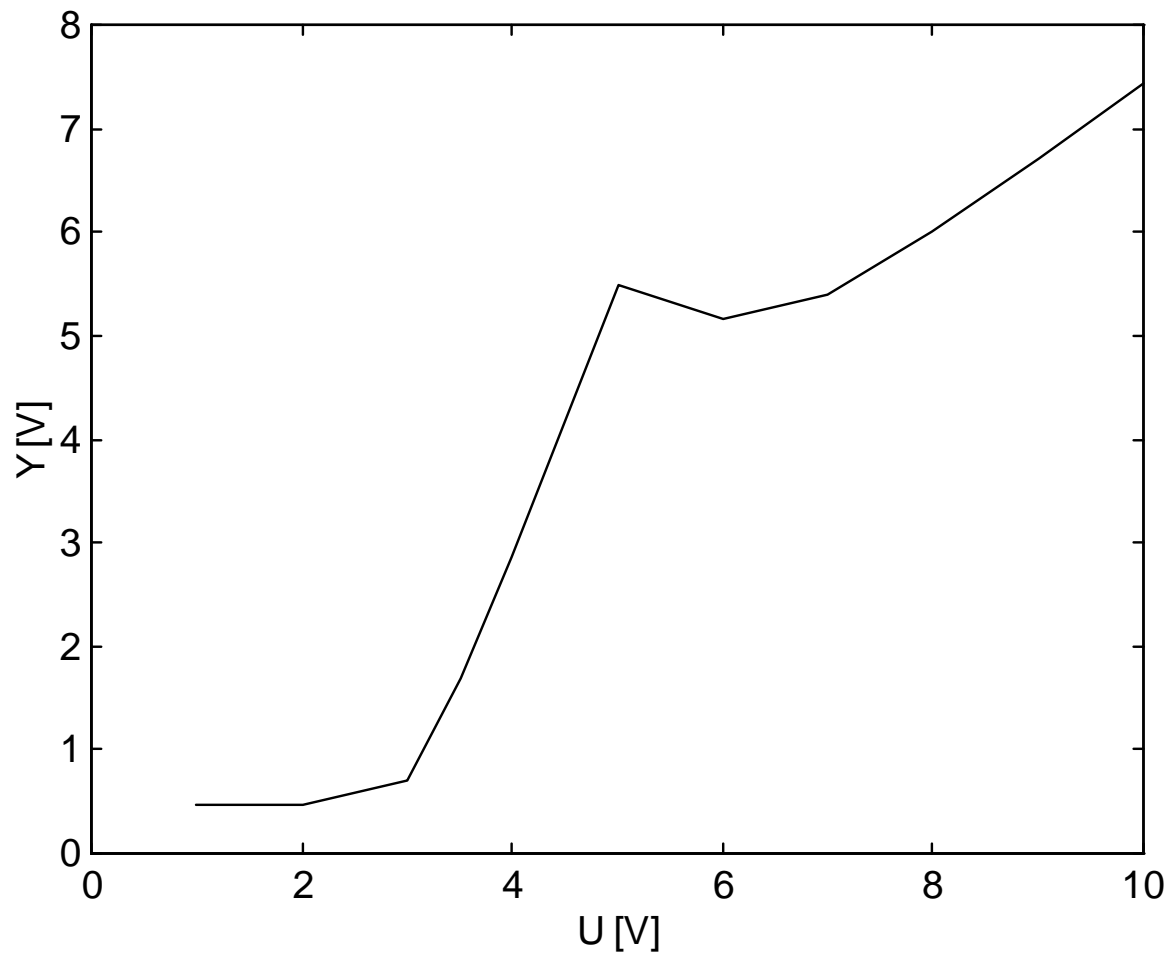


# *Motor-generator test rig*

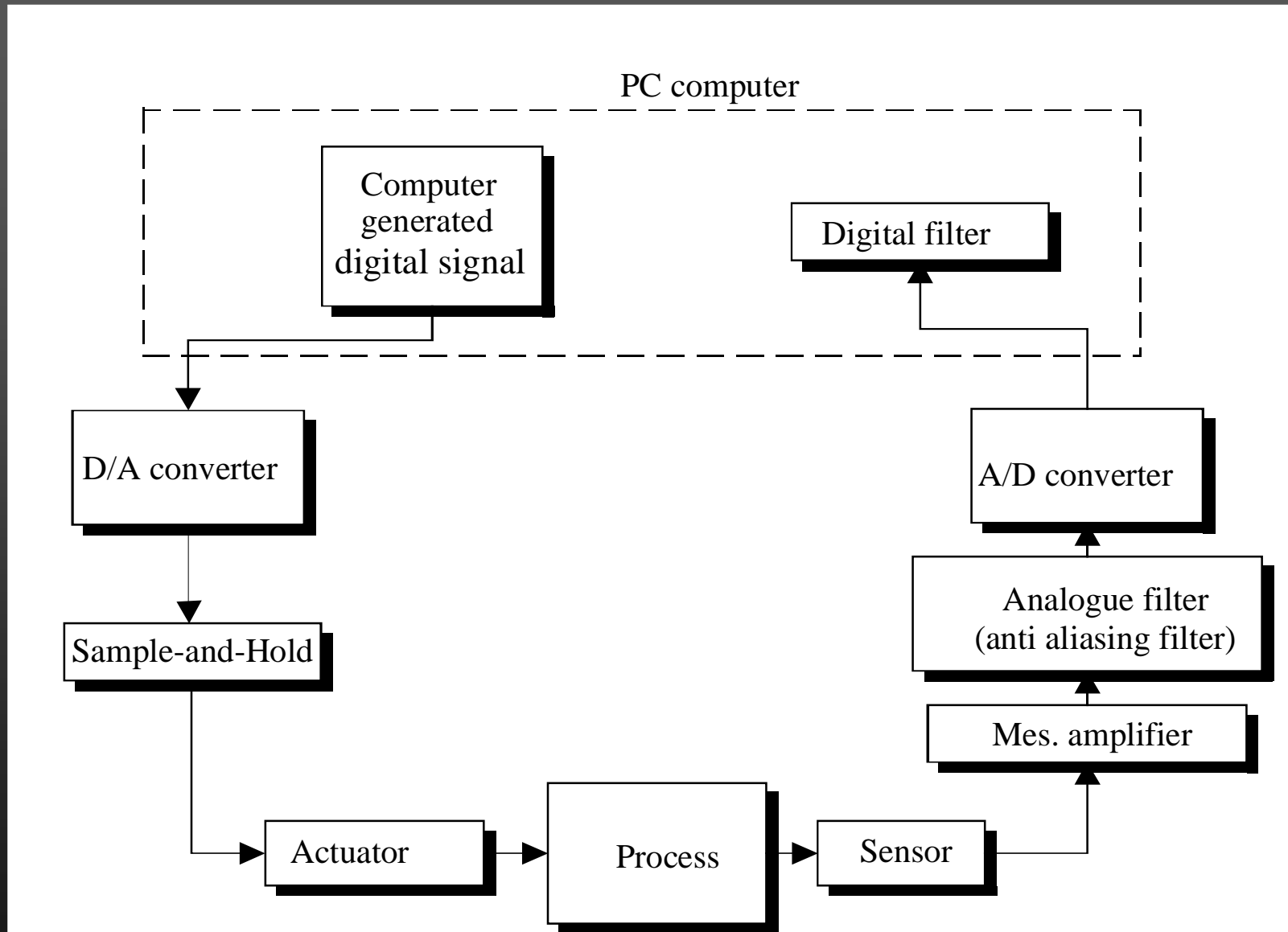
- 1 *100 W DC motor, outside excited DC generator*
- 1 *load: two 40 W bulbs*
- 1 *data acquisition with A/D converter, input signals via D/A converter and thyristor transmitter*
- 1 *nonlinear static characteristic*
- 1 *process input: voltage on motor input*
- 1 *process output: motor and generator rotations per minute*



# *Motor-generator test rig*



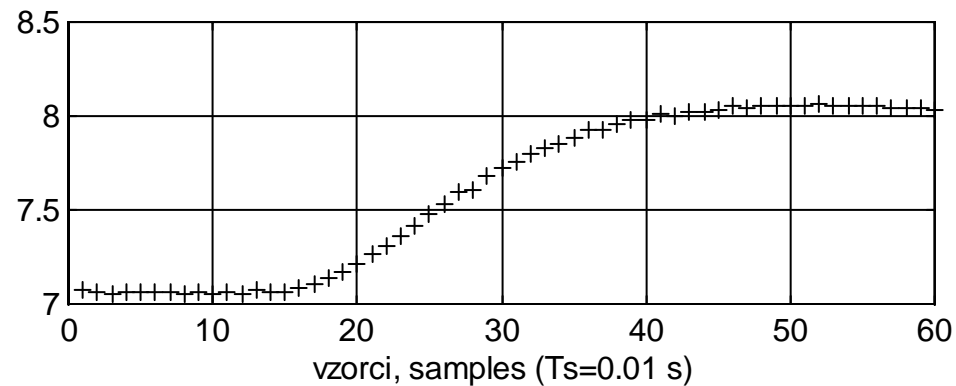
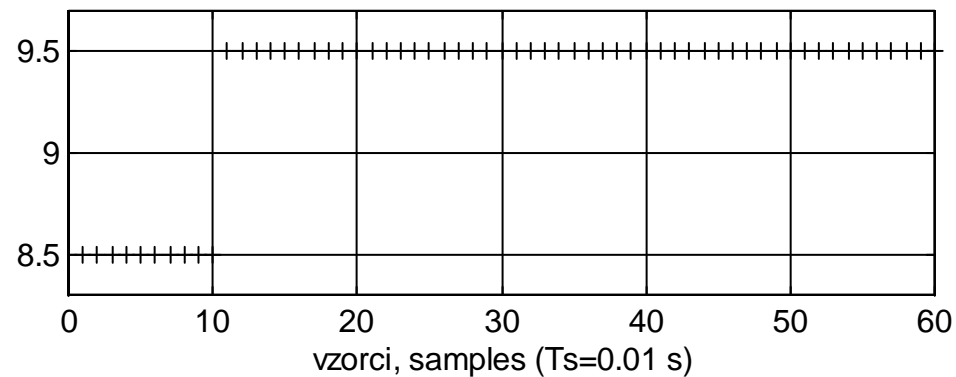
# Data acquisition





# Sampling time, input signal, pre-processing

## 1 sampling time selection



$$T_s = 0.02s$$

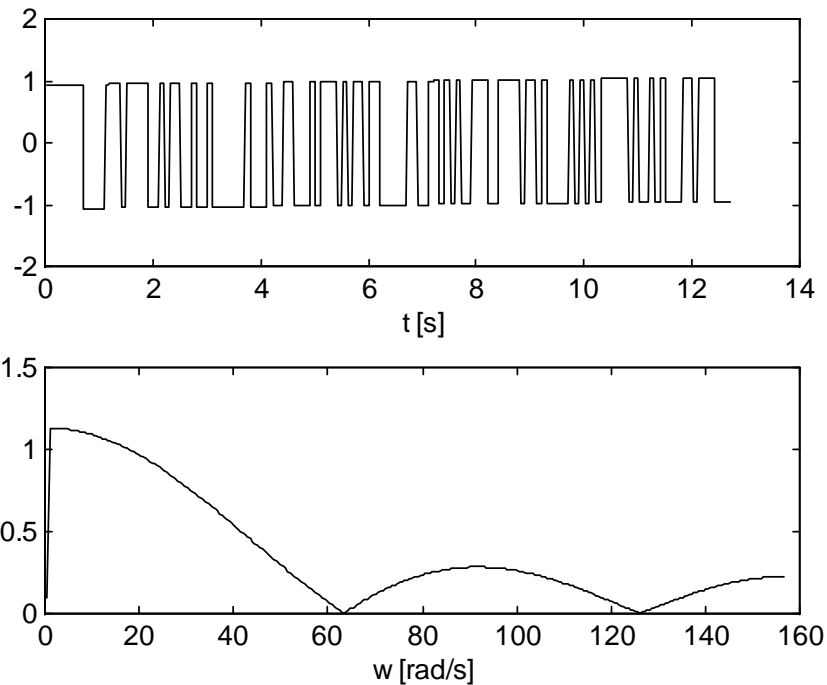
$$f_{\max} = \frac{1}{2T_s} = 25\text{Hz}$$

$$\omega_{\max} = 2\pi \cdot f_{\max} = 157\text{rad/s}$$



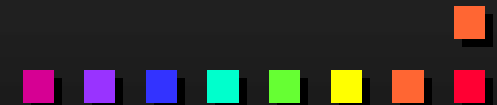
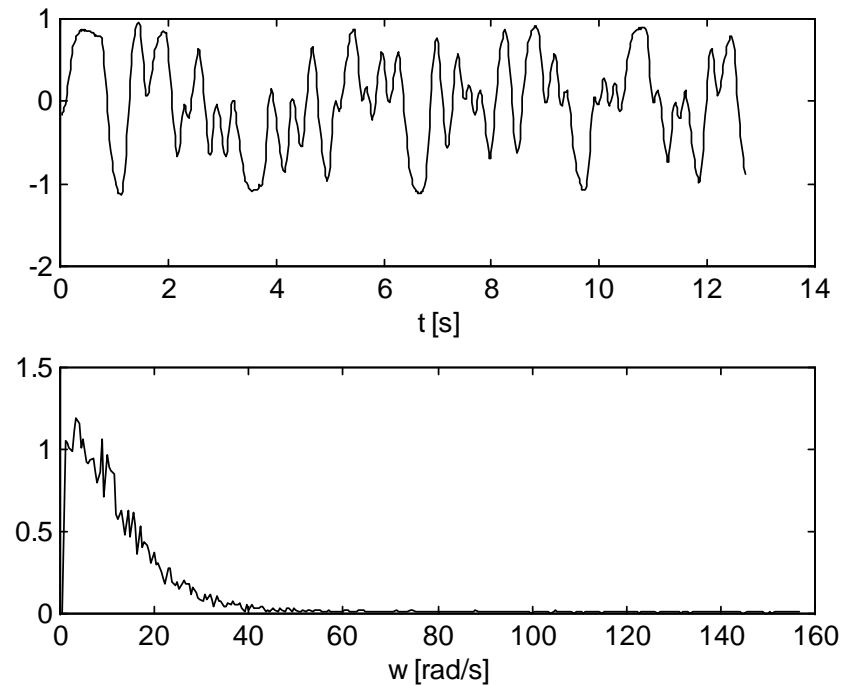
# Sampling time, input signal, pre-processing

## 1 input signal selection



## 1 PRBS signal, 10-90% settling time

## 1 removing offset and filtering



# Model structure selection

- 1 analysis of information matrix

$$\det\left(\frac{[\Psi^T \Psi]^{-1}}{N}\right)$$

$$G(j\omega) = \frac{\Phi_{uy}(j\omega)}{\Phi_{uu}(j\omega)}$$
$$W_3(j\omega) = \frac{\pi \sin(\omega T_M)}{\omega(\pi^2 - \omega^2 T_M)}$$



# *Model structure selection*

## *1 Order and regressors*

*1 2nd order*

*1 Regressors:  $y(k-1), y(k-2), u(k-1), u(k-2)$*

*1  $y(k) = -a_1 y(k-1) - a_2 y(k-2) + b_1 u(k-1) + b_2 u(k-2)$*

*1  $\begin{bmatrix} y(3) \end{bmatrix} = \begin{bmatrix} -y(2) & -y(1) & u(2) & u(1) \end{bmatrix} \begin{bmatrix} a_1 \end{bmatrix}$*

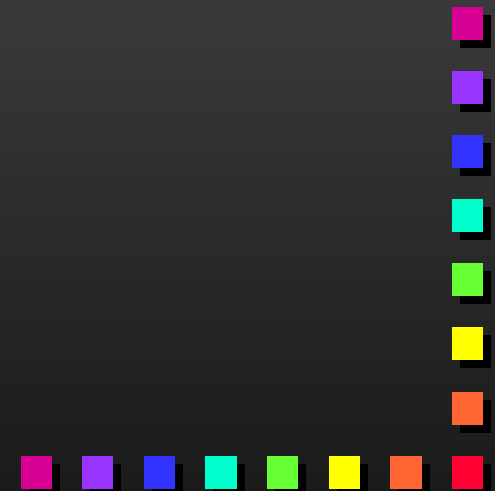
*$\begin{bmatrix} y(4) \end{bmatrix} \quad \begin{bmatrix} -y(3) & -y(2) & u(3) & u(2) \end{bmatrix} \quad \begin{bmatrix} a_2 \end{bmatrix}$*

*...*

*...*

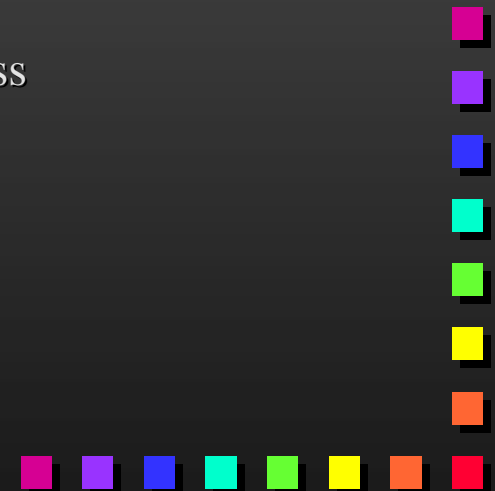
*...*

*1 Least-squares cost function - ARX*

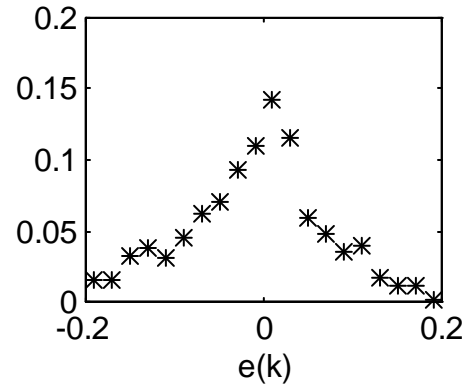
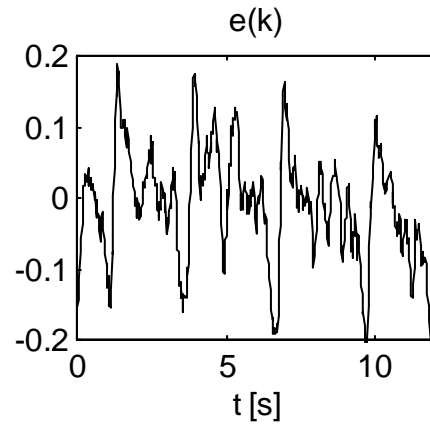


# *Model validation*

- 1 same input signal as for identification
- 1 different input signal for validation – cross-validation
- 1 test of residuals
- 1 frequency response consistency
- 1 input-output consistency
- 1 consistency of parameters, covariance matrix, parameter variances, model reduction, model purposiveness

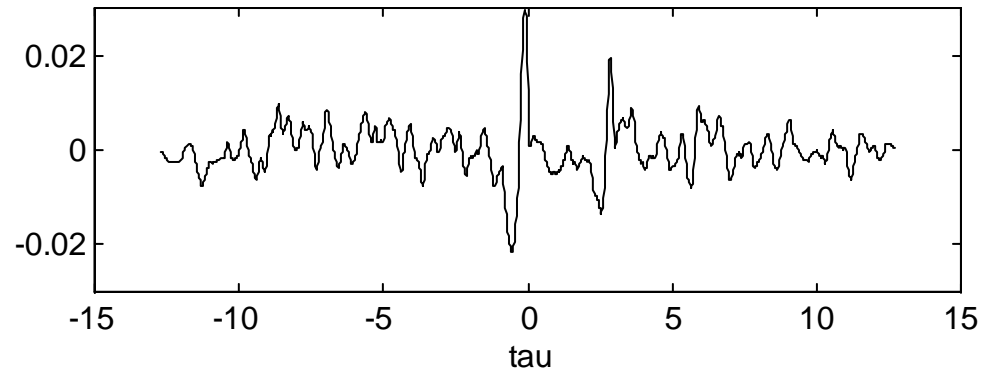
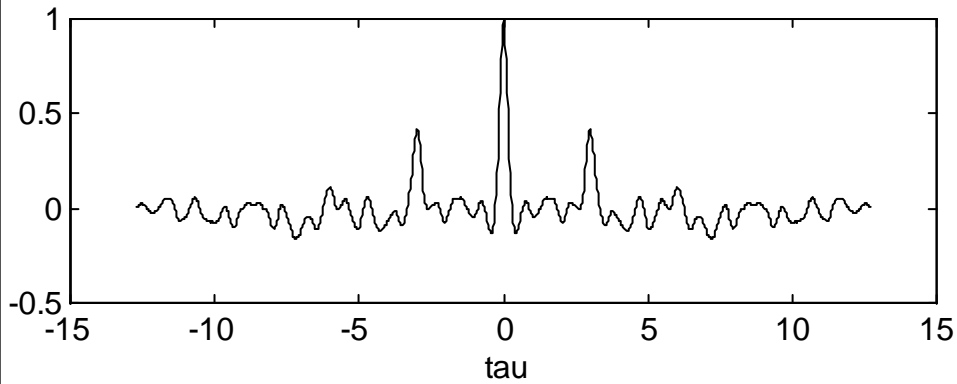


# *Test of residuals*



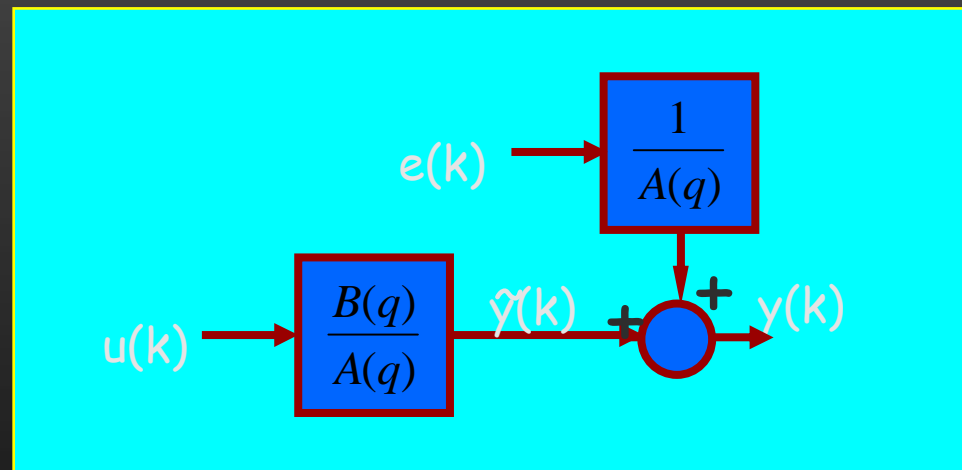
$R_{ee}$

$\phi_{eu}$

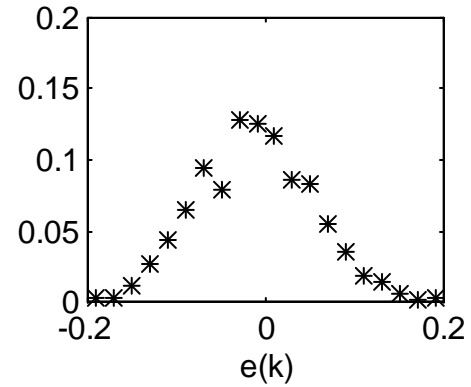
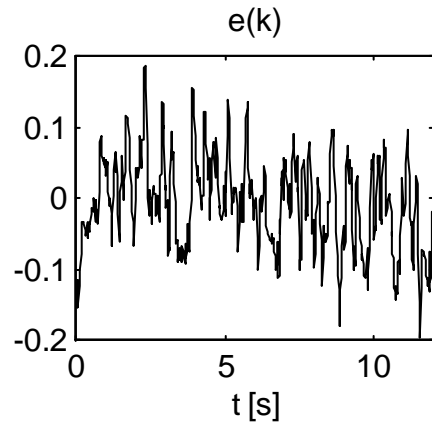


1 Autoregressive model with exogenous inputs (ARX)

$$y(k) = a_1 y(k-1) + a_2 y(k-2) + b_1 u(k-1) + b_2 u(k-2) + e(k)$$

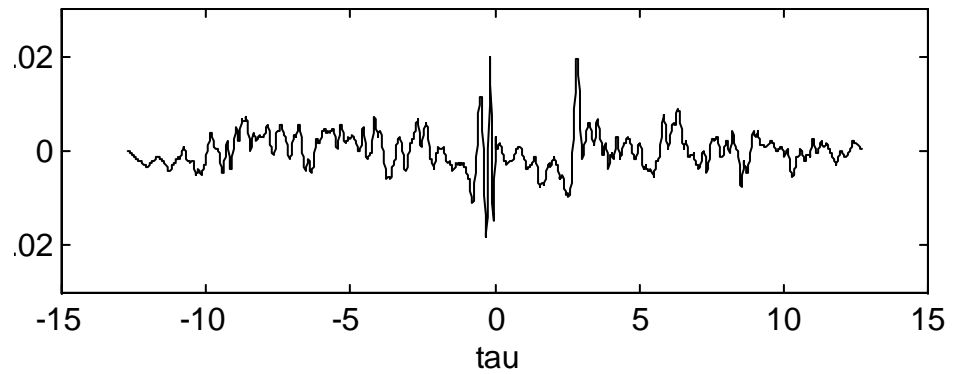
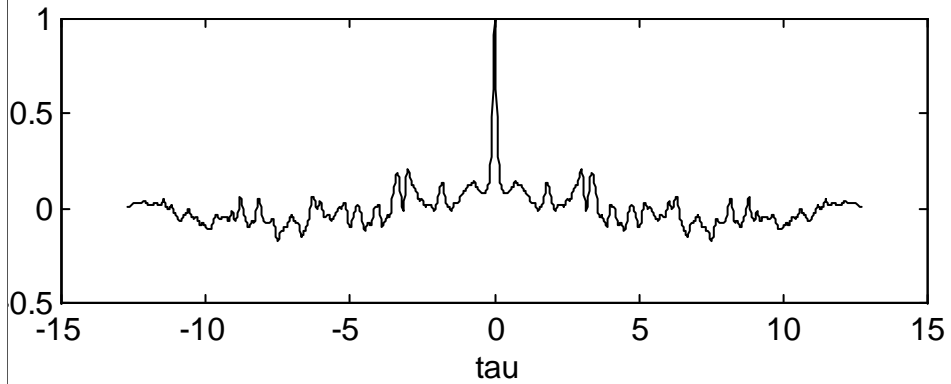


# *Test of residuals*



$R_{ee}$

$\phi_{eu}$

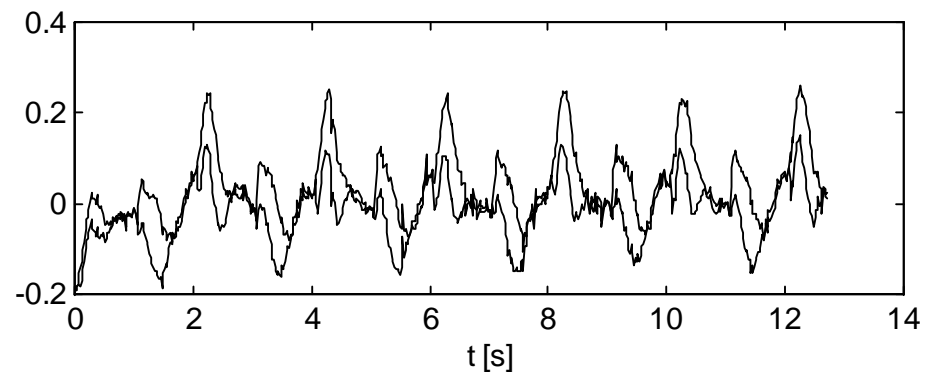
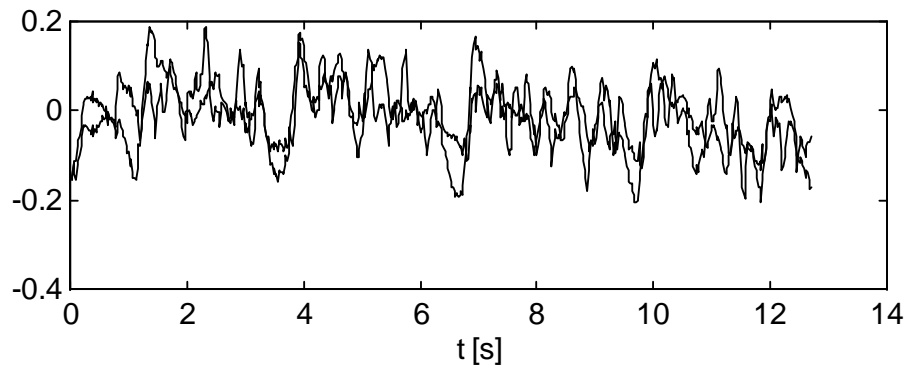
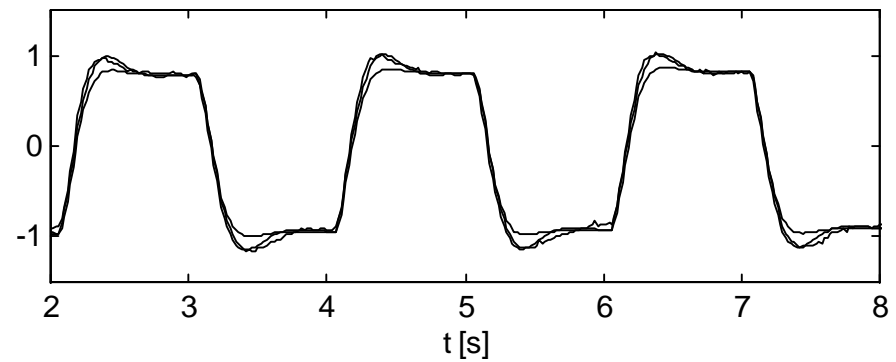
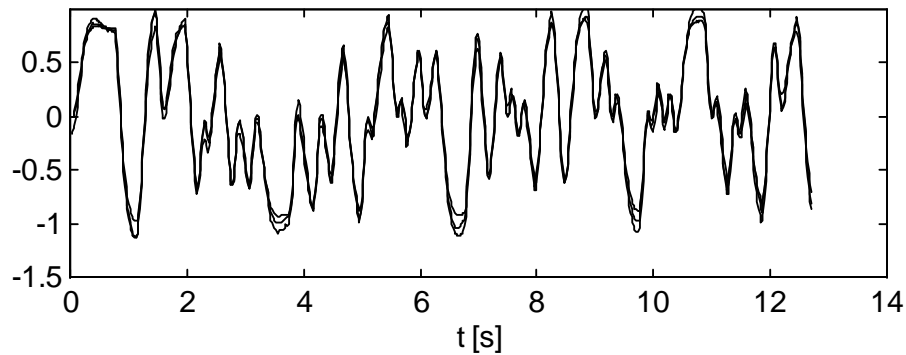




# *Input-output consistency*

Identification signal

Validation signal



# Consistency of parameters, model reduction, covariance matrix, model purposiveness

$$\text{cov}[\hat{\Theta} - \Theta] = \sigma_e^2 E\{[\Psi^T \Psi]^{-1}\}$$

$\hat{G}_{pn}$  :

$$a_1 = -1.6379; \sigma = 0.0117;$$

$$a_2 = 0.6890; \sigma = 0.0112;$$

$$b_1 = 0.0281; \sigma = 0.0013;$$

$$b_2 = 0.0165; \sigma = 0.0017;$$

$\hat{G}_{pf}$  :

$$a_1 = -1.7558; \sigma = 0.0040;$$

$$a_2 = 0.7923; \sigma = 0.0035;$$

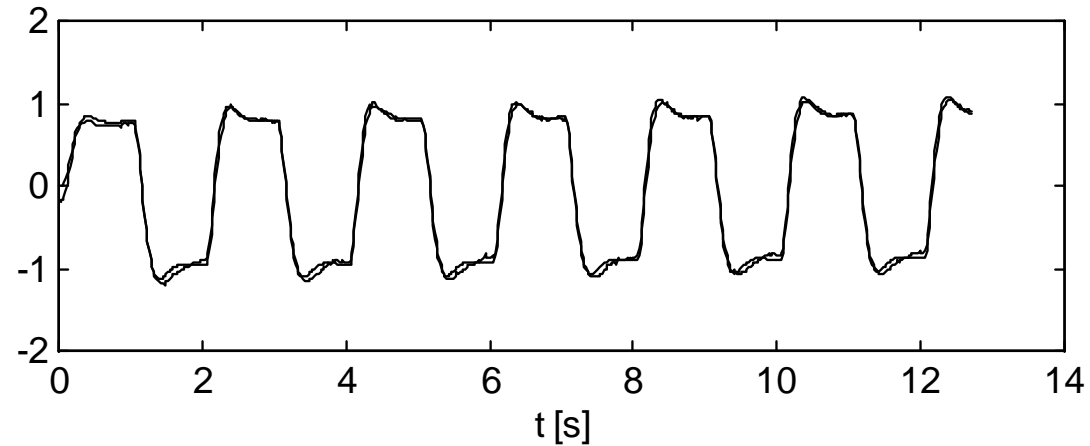
$$b_1 = 0.0590; \sigma = 0.0014;$$

$$b_2 = -0.0272; \sigma = 0.0019;$$

$$d = 2 \text{ samples}$$



# Results of identification



$\hat{G}_{final}$  :

$$a_1 = -1.7144;$$

$$a_2 = 0.7589;$$

$$b_1 = 0;$$

$$b_2 = 0.0390;$$

$$d = 2 \text{ samples};$$



- 1 Dynamic systems identification is a routine operation in practice whenever we deal with linear and close-to-linear systems.
- 1 Importance of input signal selection, sampling time selection, signals pre-processing, selection of regressors and model validation.
- 1 System identification is highly ITERATIVE, but can never be automatic.
- 1 The result of modelling is always a model, never a copy.
- 1 The quality of model depends on the entire identification process.



# *MATLAB System Identification Toolbox*

- 1 ARX, ARMAX, OE, BJ, ARARX, ARARMAX, PEM, IV
- 1 Recursive models: ARMAX, ARX, BJ, OE, PEM
- 1 Time-series identification: AR, ARMA, IVAR
- 1 Segmentation of data
- 1 Structure search
- 1 Other tools

