

# System Identification and Loop Shaping

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Departamento de Ingeniería Eléctrica



**IEE2683 – Laboratorio de Control Automático**

## System Identification

- Preliminaries

- Non-parametric identification in the frequency domain

- Non-parametric closed-loop identification in the frequency domain

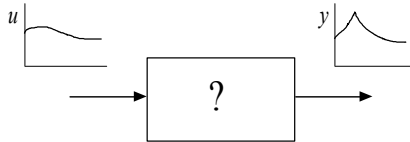
- Parametric Identification

## Loop-Shaping

- Open-loop vs closed-loop specifications

- Compensators

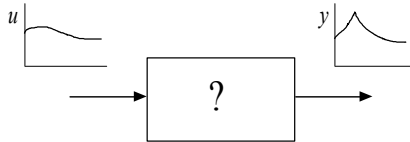
## Problem setup



### Elements

1. Unknown process with at least one input and one output
2. Actuator to introduce an user-designed input
3. Sensor to measure corresponding output

## Problem setup



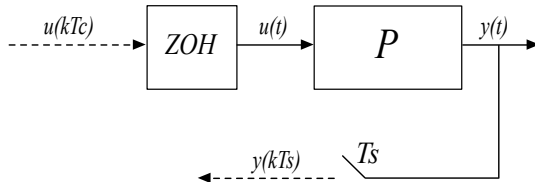
### Elements

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### Goals

1. Determine a model using input/output experimental data
2. In the form of a set of equations
3. Or in the form of a characteristic response

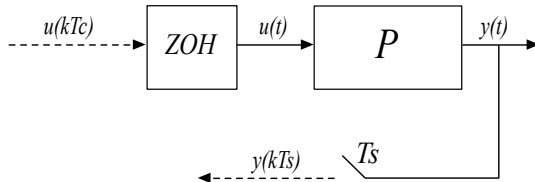
## Problem setup



### Assumptions

1. Continuous-time LTI and (stable) process
2. Arbitrary sampling frequency for sensor and actuator
3. Strictly proper transfer function without non-minimum phase zeros

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### Challenges

1. Design a set of inputs  $u$  to characterize  $P$  based on outputs  $y$
2. Design a processing routine to characterize  $P$  based on input/output data

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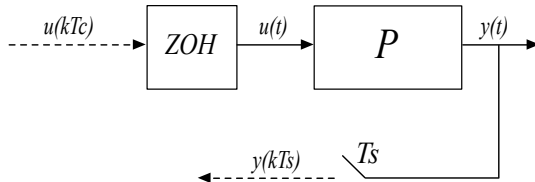
Parametric Identification

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### Goals

1. Obtain an estimate of the frequency response  $\hat{G}(e^{j\omega})$
2. Obtain the coherence spectrum  $\kappa$



## Input selection

### Typical signals

1. Zero function
2. Step function
3. Sinusoidal
4. Periodic signals: sum of sinusoids
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### Winners

1. Periodic function: gives estimate of bandwidth
2. PRBS of period  $M$ : p.e. of order  $M$  with white noise-like spectrum

# Filtering

## Artifacts

1. Drifts
2. Grid noise
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2. Time-domain average exploiting periodic nature of input:
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  - 2.2 Perform direct averaging to obtain smoothed data
3. Smoothing in the frequency domain (next slide)

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## Input design revisited

1. Construct a PRBS of length  $M$
2. Append several periods to obtain input

# Spectral leakage

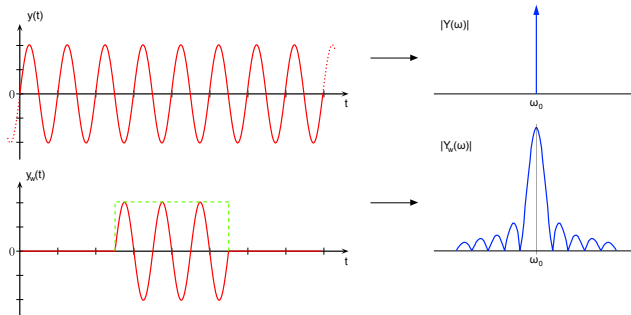
## Problem

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## Spectral leakage

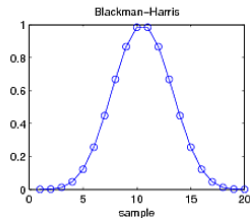
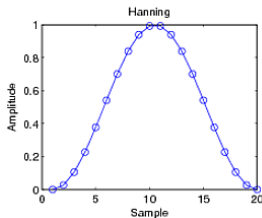
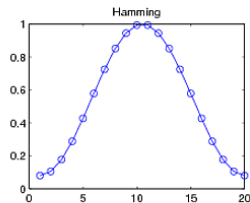
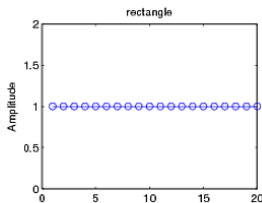
### Solution

Windowing of signals: non-square window function to improve spectral properties

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# Processing

On a windowed vector of identification data.

## Correlation function

$$\hat{R}_u^N(\tau) = \frac{1}{N} \sum_{t=1}^N u(t)u(t-\tau), \quad \tau \in [-\gamma, \gamma] \quad (1)$$

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Calculate estimations for  $\hat{R}_{yu}^N$ ,  $\hat{\Phi}_{yu}^N$ ,  $\hat{R}_y^N$ , and  $\hat{\Phi}_y^N$

## Processing

### Smoothed Estimate of Frequency Response

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### Coherence Spectrum

$$\kappa = \sqrt{\frac{|\hat{\Phi}_{yu}^N(\omega)|^2}{\hat{\Phi}_y^N(\omega)\hat{\Phi}_u^N(\omega)}} \quad (5)$$



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**Non-parametric closed-loop identification in the frequency domain**

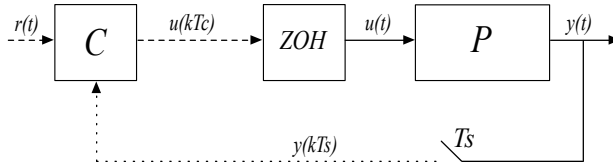
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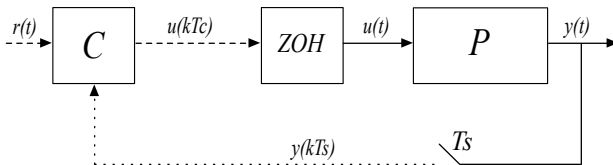
Open-loop vs closed-loop specifications

Compensators

## Problem setup



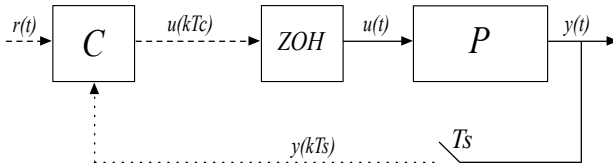
## Problem setup



### Goals

1. Obtain an estimate of the frequency response  $\hat{G}(e^{j\omega})$
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## Problem setup



### Problem

1. Correlation between inputs and outputs due to feedback !!
2. Controller tracks the set-point: no p.e.

## Problem setup

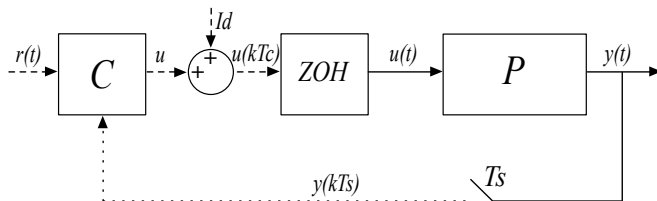
### Solution

1. Use a “bad” controller: only stabilization
2. Add an exiting input after the controller
3. Use direct input to the plant to perform identification

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# Overview

## Methods

1. Least-squares
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2. Weighted least-squares
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## Structures

1. ARX
2. ARMAX
3. ARIMAX
4. NARMAX
5. ...

# Overview

## Subspace methods in the time domain

1. Give a state-space representation of the system
2. Less sensitive to measurement noise
3. Complicated algorithms
4. Often give unstable models
5. See N4SID (exists in Matlab)

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## Subspace methods in the time domain

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## Subspace methods in the frequency domain

1. Give a state-space representation of the system
2. Less sensitive to measurement noise
3. Complicated algorithms using samples of the frequency response
4. Toolboxes for Matlab exist

## System Identification

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Non-parametric identification in the frequency domain

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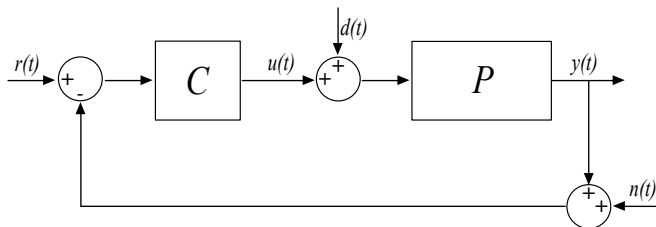
Parametric Identification

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# Overview



## Main ideas

1. Look to design a controller  $C$  to satisfy closed-loop specifications
2. Convert closed-loop specifications into constraints on the open-loop gain  $G_0 = CP$
3. Obtain the plant Bode diagram and shape  $G_0$  starting from  $C = 1$
4. "Add" compensators until desired shape is achieved

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## Specifications

### Stability

Given cross-over frequency  $w_c$  the phase must be above -180 degrees. The distance between the phase of  $G_0(j\omega)$  and -180 degrees is called the phase-margin.

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Larger phase margins generally corresponds to a smaller overshoot for the step response of the closed-loop system.



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Larger phase margins generally corresponds to a smaller overshoot for the step response of the closed-loop system.

## Reference tracking

A sufficient condition to track a reference with an error at least  $k_t \ll 1$  times smaller than the reference over a frequency range  $[0, \omega_t]$  is given by:

$$|G_0(j\omega)| \geq \frac{1}{k_t} + 1, \quad \forall \omega \in [0, \omega_t] \quad (6)$$

# Specifications

## Disturbance rejection

A sufficient condition to attenuate a disturbance at the output at least  $k_d \ll 1$  times over a frequency range  $[0, \omega_d]$  is given by:

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## Noise rejection

A sufficient condition to attenuate the measurement noise at the output at least  $k_n \ll 1$  times over a frequency range  $[\omega_n, \infty)$  is given by:

$$|G_0(j\omega)| \leq \frac{k_n}{k_n + 1}, \quad \forall \omega \in [\omega_n, \infty) \quad (8)$$

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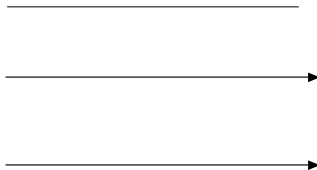
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# Compensators

P gain

$$C_p = K$$



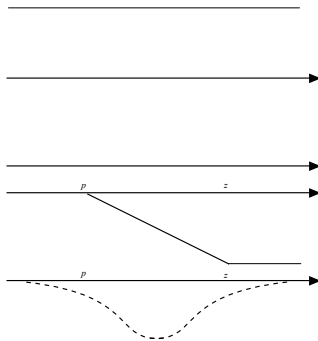
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$$C_{lag} = \frac{s/z + 1}{s/p + 1}, \quad p < z$$



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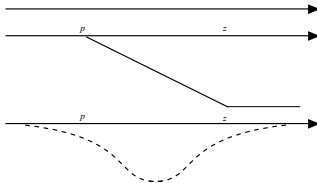
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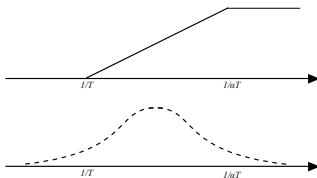
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$$C_{lag} = \frac{s/z + 1}{s/p + 1}, \quad p < z$$



## Lead compensator

$$C_{lead} = \frac{Ts + 1}{\alpha Ts + 1}, \quad \alpha < 1$$



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