



Chapter 1. Introduction to Intelligent Control System

Luis Miguel Bergasa

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Introduction to ICS General information

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 - Eduardo Romera (eduardo.romera@edu.uah.es)
- Degree in
 - Computer Science Engineering (Obligatory)
 - > Computer Engineering, Engineering in Information Systems (Optional)
- Course Webpage: http://www.uah.es/aula_virtual/

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Computer Science Engineering

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HORA	LUNES	MARTES	MIÉRCOLES	JUEVES	VIERNES
8:00/8:55		Lab. Fundamentos de la Ciencia de Datos – A1 (NL8)			
9:00/9:55		Lab. Fundamentos de la Ciencia de Datos – A1 (NL8)		OPT: ARQ. Y DISENO DE SISTEMAS WEB Y C/S (NA6)	
10:00/10:55	SISTEMAS DE CONTROL INTELIGENTES (NA6)	FUNDAMENTOS DE LA CIENCIA DE DATOS (SA8) (47)	Lab. Aplicaciones de Soyf Computing (SLT) (£F) Lab. Sistemas Audiovisualies y App. Mul. (SLT) OPT: REDES DEFINIDAS POR SOFTWARE (EA2) (£F)	Lab. Arq. y Diseno de Sistemas Web y C/s – A1 (NL11) OPT: SIST. CONTROL PARA ROBOTS (NA6)	Lab. Sistemas Audiovisualies y App. Mul. (SL7)
11:00/11:55	SISTEMAS DE CONTROL INTELIGENTES (NA6)	FUNDAMENTOS DE LA CIENCIA DE DATOS (SA8) (47)	Lab. Aplicaciones de Sotf Computing (SLT) (LD) Lab. Sistemas Audiovisualies y App. Mul. (SLT) OPT: REDES DEFINIDAS POR SOFTWARE (EAZ) (LT)	Lab. Arq. y Diseno de Sistemas Web y C/s - Al (NLII) OPT: SIST. CONTROL PARA ROBOTS (NA6)	Lab. Sistemas Audiovisualies y App. Mul. (SL7)
12:00/12:55	Lab. Sistemas de Control Inteligentes – A1 (OL1) -A2 (OL5)	Lab. Fundamentos de la Ciencia de Datos – A2 (NL8)	Lab. Aplicaciones de Sotf Computing (SL7) Lab. Redes Definidas por Software (EL11)	Lab. Sist. Control Para Robots (PL21) Lab. Arq. y Diseno de Sistemas Web y C/s – A2 (NL11)	Lab. Sistemas Audiovisualies y App. Mul. (SL7)
13:00/13:55	Lab. Sistemas de Control Inteligentes – A1 (OL1) -A2 (OL5)	Lab. Fundamentos de la Ciencia de Datos – A2 (NL8)	Lab. Aplicaciones de Sotf Computing (SL7) Lab. Redes Definidas por Software (EL11)	Lab. Sist. Control Para Robots (PL21) Lab. Arq. y Diseno de Sistemas Web y C/s – A2 (NL11)	Lab. Sistemas Audiovisualies y App. Mul. (SL7)
14:00/14:55					
15:00/15:55	Lab. Sistemas de Control Inteligente – A3 (OL5)	Lab. Sistemas de Visión Artificial (OL11) Lab. Patrones Software –A1 (NL11)		Lab. Sistemas de Visión Artificial (OL11) Lab. Arq. y Diseno de Sistemas Web y C/s – A3 (NL10)	
16:00/16:55	Lab. Sistemas de Control Inteligente – A3 (OL5)	Lab. Sistemas de Visión Artificial (OL11) Lab. Patrones Software –Al (NL11)		Lab. Sistemas de Visión Artificial (OL11) Lab. Arq. y Diseno de Sistemas Web y C/s – A3 (NL10)	
17:00/17:55		OPT: PATRONES SOFTWARE (OA3) (#)			
18:00/18:55		OPT: PATRONES SOFTWARE (OA3) (#)			
19:00/19:55		Lab. Patrones Software -A2 (NL9)			
20:00/20:55		Lab. Patrones Software -A2 (NL9)			

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Computer Engineering, Engineering in Information Systems

HORA	LUNES	MARTES	MIÉRCOLES	JUEVES	VIERNES
8:00/8:55					
9:00/9:55				OPT: ARQ. Y DISENO DE SISTEMAS	
			Lab. Aplicaciones de Sotf Computing (SL7)	WEB Y C/S (NA6)	
10:00/10:55		GESTION DE REDES Y SEGURIDAD (NA6) (P)	Lab. Sistemas Audiovisualies y App. Mul. (SLT) SISTEMAS DE CONTROL INTELIGENTES (NA8) OPT: REDES DEFINIDAS POR SOFTWARE (EA2) (E)	Lab. Arg. y Diseno de Sistemas Web y C/s – A1 (NL11) OPT: SIST. CONTROL PARA ROBOTS (NA6)	Lab. Sistemas Audiovisualies y App. Mul. (SL7) OPT: COMPUTACIÓN DE ALTAS PRESTACINES (SA5A)
11:00/11:55		GESTION DE REDES Y SEGURIDAD (NA6) (#)	Lab. Aplicaciones de Sosf Computing (SLT) (D) Lab. Sistemas Audiovisualies y App. Mul. (SLT) SISTEMAS DE CONTROL INTELIGENTES (NAS) OPT: REDES DEFINIDAS POR SOFTWARE (EA.2) (D)	Lab. Arq. y Diseno de Sistemas Web y C/s – Al (KLII) OPT: SIST. CONTROL PARA ROBOTS (NA6)	Lab. Sistemas Audiovisualies y App. Mul. (SL7) OPT: COMPUTACIÓN DE ALTAS PRESTACINES (SA5A)
12:00/12:55		Lab. Gestion de Redes y Seguridad- A1 (NL5) -A2 (EL11)	Lab. Aplicaciones de Sotf Computing (SL7) Lab. Sistemas de Control Inteligentes (OL1) Lab. Redes Definidas por Software (EL11)	Lab. Sist. Control Para Robots (PL21) Lab. Arq. y Diseno de Sistemas Web y C/s – A2 (NL11)	Lab. Sistemas Audiovisualies y App. Mul. (SLT) Lab. Computación De Altas Prestacines (NLS)
13:00/13:55		Lab. Gestion de Redes y Seguridad- A1 (NL5) -A2 (EL11)	Lab. Aplicaciones de Sotf Computing (SL7) Lab. Sistemas de Control Inteligentes (OL1) Lab. Redes Definidas por Software (EL11)	Lab. Sist. Control Para Robots (PL21) Lab. Arq. y Diseno de Sistemas Web y C/s – A2 (NL11)	Lab. Sistemas Audiovisualies y App. Mul. (SL7) Lab. Computación De Altas Prestacines (VL5)
14:00/14:55				•	
15:00/15:55		Lab. Sistemas de Visión Artificial (OL11) Lab. Patrones Software –A1 (NL11)		Lab. Sistemas de Vision Artificial (OL11) Lab. Arq. y Diseno de Sistemas Web y C/s – A3 (NL10)	
16:00/16:55		Lab. Sistemas de Visión Artificial (OL11) Lab. Patrones Software –A1 (NL11) –		Lab. Sistemas de Vision Artificial (OL11) Lab. Arq. y Diseno de Sistemas Web y C/s – A3 (NL10)	
17:00/17:55		OPT: PATRONES SOFTWARE (OA3) (#)			
18:00/18:55		OPT: PATRONES SOFTWARE (OA3) (#)			
19:00/19:55		Lab. Patrones Software -A2 (NL9)			
20:00/20:55		Lab. Patrones Software -A2 (NL9)			

Introduction to ICS Chapters

Theory

- Chapter 1. Introduction to Intelligent Control Systems
- Chapter 2. Basic concepts of discrete control
- > Chapter 3. Introduction of the Neural Networks
- > Chapter 4. Neural Control
- > Chapter 5. Introduction to the Fuzzy Logic
- Chapter 6. Fuzzy Control

Practices

- Training 1. Introduction to MATLAB-SIMULINK
- > Training 2. Identification and neural control
- > Training 3. Fuzzy control
- > Final project. ICS for a mobile robot

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Introduction to ICS Calendar Temario Presentación Tema 1 (T) Tema 2-1 (T) Tema 2-2 (T) Training MATLAB - 1 Training MATLAB - 2 Tema 2-3 (P) Tema 2-3 (P) Tema 2-4 (P) Training MATLAB - 3 Training MATLAB - 3 Training MATLAB - 4 Tema 3-1 (T) Tema 3-2 (T) Training CN - 1 Training CN - 2 Tema 3-3 (T) Tema 3-4 (T) Training CN - 3 Training CN - 3 Training CN - 4 Tema 4-1 (T) Tema 4-2 (T) Tema 4-2 (T) 09-09 10-11 1h 09-09 11-12 1h 16-09 10-12 S5-S6 16-09 12-14 2h Problemas CN-1 Problemas CN-2 Tema 5-1 (T) Tema 5-2 (T) Tema 5-2 (T) Tema 5-4 (T) Tema 6-1 (T) Tema 6-2 (T) S7-S8 23-09 10-12 2h S25-S26 21-10 12-14 2h S9-S10 23-09 12-14 2h 10-12 S27-S28 28-10 2h S11-S12 30-09 10-12 S29-S30 28-10 12-14 S13-S14 30-09 10-12 S31-S32 04-11 10-12 S15-S16 10-12 07-10 S33-S34 04-11 12-14 Problemas CB-1 Problemas CB-2 Tema 6-3 (T) Tema 6-4 (T) S17-S18 07-10 12-14 S35-S36 11-11 10-12 2h S19-S20 14-10 10-12 S37-S38 11-11 12-14 Training CB Tema 4-1 (1) Tema 4-2 (T) Training CN - 5 Training CN - 6 Tema 4-3 (T) Tema 4-4 (T) Training CB – 1 Training CB – 2 Problemas CB-3 S21-S22 14-10 12-14 S39-S40 18-11 10-12 Problemas CB-4 S23-S24 21-10 12-14 2h S41-S42 18-11 12-14 Training CB – 3 Training CB – 4 2h TCB Test de conceptos básicos Miniproyecto S44-S46 25-11 S47-S50 02-12 10-14 Miniproyecto S51-54 16-12 10-14 Miniproyecto

Introduction to ICS Evaluation

Writing test 40 points

Research final project 40 points

Report quality
Scientific method and results
Presentation
20%

Lab Training 20 points

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Introduction to ICS Bibliography

Basic

➤ A First Course in Fuzzy and Neural Control. Hung T. Nguyen, Nadipuram R. Prasad, Carol L. Walker, Elbert A. Walker. Chapman and Hall/CRC. 2002

Complementary

- Neural Network Design. Martin T. Hagan, Howard. B. Demuth, Mark Beale. PWS Publishing Company, Thomson Learning. 2002
- Application of Neural Networks to Adaptive Control of Nonlinear Systems. G. W. Ng. Research Studies Press Ltd. England. 2003
- Neural Control Engineering: The Emerging Intersection between Control Theory and Neuroscience. Steven J. Schiff. The MIT Press. Firts edition 2012
- Redes Neuronales y Sistemas Borrosos. Bonifacio Martín del Brío y Alfredo Sanz Molina. Ed. RAMA. 2001
- Fuzzy Control. Kevin M. Passino, Stephen Yurkovich. Addison Wesley Publishing Company. First edition 1997
- An Introduction to Fuzzy Control. D. Driankov, H. Hellendoorn and M. Reinfrank. Springer-Verlag. 1996
- Adaptive Fuzzy Systems and Control Design and Stability Analysis. Li-Xin Wang. University of California at Berkeley. PTR Prentice Hall. 1994
- Controladores en lógica borrosa. Javier Holgado Corrales. Servicio de Publicaciones de la Universidad de Cadiz. 1995

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Introduction to ICS Course Requirements

- Essential prerequisites
 - Classical Control
 - Linear algebra
 - Numeric methods
 - Basic knowledge of MATLAB
- Will not be covered
 - Review of all architectures and learning rules (only focused in control systems)
 - > HW implementation
 - Biology

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Introduction to ICS General Goals

- This course gives an introduction to discrete control, neural network architectures and fuzzy logic principles
- Emphasis is placed on their application to control systems from a theoretical and practical point of view
- An implementation of a neural controller, fuzzy controller and neuro-fuzzy controller is done
- ➤ A review of the genetic algorithms and their application in control systems is then carried out

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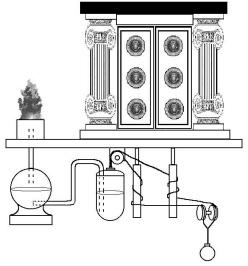
Introduction to ICS General Overview

- Intelligent Control Systems
 - Neural Control Systems (NCS)
 - Fuzzy Control Systems (FCS)
 - Neuro-Fuzzy Control Systems
- Soft Computing
 - Neural Networks (NNs)
 - Fuzzy Logic (FL)
 - Genetic Algorithms (GA)

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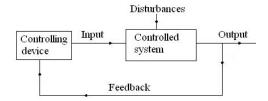


Amazing control system invented about 2000 years ago by Hero of Alexandria (c. I AD)

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A Prelude to Control Theory Examples of Control Problems

- The goal of control theory is to force a system to behave the way we want
- Control system:



- Classification:
 - Open-loop control
 - Closed-loop control

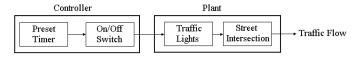
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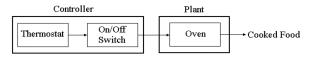
A Prelude to Control Theory Examples of Control Problems

Open-loop control

- Control systems operating without feedback regarding the system behavior
- Example 1 (Traffic light)



> Example 2 (Conventional oven)

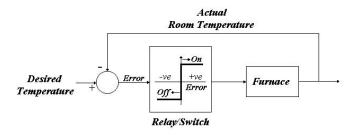


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A Prelude to Control Theory Examples of Control Problems

Closed-loop or feedback control

- ➤ The behavior of the system is observed by some sensory device, and the observations are fed back so that a comparison can be made about how well the system is behaving in relation to some desired performance.
- > Example 1 (Temperature control)

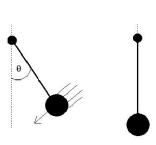


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A Prelude to Control Theory Stable and Unstable Systems

> Stability of a system

> The ability of the system to maintain equilibrium or resume its original position after displacement.







>Unstable system

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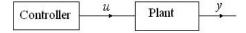
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A Prelude to Control Theory

A look at controller design

Control law

> The engineering problem: How do you find the function u and how do you implement it?



Example: cruise control (open-loop case)

We want to keep the speed of a car at y₀ = 65 mph for all t > t₀

$$y(t) = f(u(t))$$

Given y_0 , the problem is to find the control function $u_0(t)$ such that

$$f(u_0(t)) = y_0$$
 for $t > t_0$

f -> mathematical model of the plant

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A Prelude to Control Theory A look at controller design

Classical approach: standard control theory

- Focuses on finding a mathematical model of the plant
 - > Analysis, collecting information pertinent to the control problem
 - > Synthesis, constructing a successful control law

> Problems:

- > Requires detailed knowledge of the plant
- ➤ The model is, at best, an approximate representation of the actual physical system

$$\frac{d^2x(t)}{dt^2} + \frac{a\,dx(t)}{dt} = b\,u(t) \quad ,x(t): car's \ position$$

$$y(t) = \frac{dx(t)}{dt}$$
, $y(t)$: car's velocity $\longrightarrow \frac{dy(t)}{dt} + ay(t) = bu(t)$

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A Prelude to Control Theory

A look at controller design

- Classical approach: standard control theory
 - Concrete solution

$$y(t) = y_0, for t > t_0 \rightarrow \frac{dy(t)}{dt} = 0 \rightarrow u(t) = \left(\frac{a}{b}\right) y_0$$

Second-order linear differential equation: Laplace Transforms in the frequency domain

$$\frac{dy(t)}{dt} + a y(t) = b u(t)$$

$$\downarrow L\{\}$$

$$sY(s) + aY(s) = bU(s) \rightarrow U(s) = \frac{(s+a)}{b}Y(s)$$

$$\downarrow L^{-1}\{\}$$

$$u(t)$$

$$Y(s) = L\{y(t)\}$$

$$U(s) = L\{u(t)\}$$

$$y(t) = L^{-1}\{Y(s)\}$$

$$u(t) = L^{-1}\{U(s)\}$$

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A Prelude to Control Theory A look at controller design

- > Classical approach: standard control theory
 - Closed-loop control

$$y(t)-y_0 = e(t) \rightarrow u(t) = g(e(t)) = h(y(t), y_0)$$

- The problem is to find the function g or to approximate it from observable values of u(t) and y(t)
- For LTI systems the so-called proportional integral derivative (PID) controllers work satisfactorily
- There exist systematic ways to design successful PID controllers knowing the plant model.

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A Prelude to Control Theory A look at controller design

Modern approach: intelligent control theory

- How can you control complicated systems whose plant dynamics are difficult to know?
- We can approximate a plausible control law, either from a collection of if ... then. . . rules or from training data without a mathematical model for the plant
- ➤ This is the philosophy of the fuzzy and neural approaches to control (weaker knowledge)
- ➤ There are parameters in these controllers but they are adjusted by training samples or trial and error. There is no need for analytical mathematical models in this process.

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