# Autonomic Computer Systems CS321: Introduction to Control Theory

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Prof. Dr. Christian Tschudin
Computer Science Department, University of Basel

# Why Control Theory (in a network centric course)?

"Dynamics" usually a second thought in CS: functionality first. However, performance (beyond speed) matters in practice: unstable operation is useless.

- Dynamics is key for TCP
- Control Theory for engineering (using computer programs):
  - motor control of an inverted pendulum, e.g. Segway
  - Mindstorm version: see

http://www.brumann-art-production.ch/NXT\_Standalone/

Can control theory be applied in a (autonomic) computing center?

#### Overview 2011-12-01

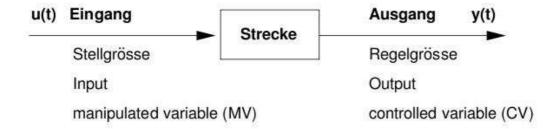
- Introduction to Control Theory problem setting, history, modeling
- Hellerstein et al slide set

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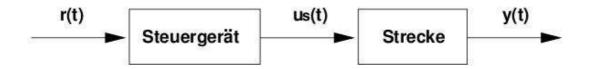
# **Terminology**

- Control theory Theorie der Regelsysteme
- Plant, or process Strecke
  - "the thing to control"
  - has inputs: actuator, manipulated variable
  - has outputs: sensor, controlled variable



# a) Feedforward Control (Steuerung)

We "prepend" a controlling device (controller) that, given some reference value r, generates the appropriate control signal u such that  $r \approx y$ .



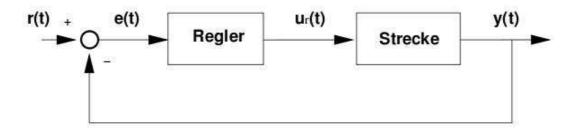
This is also called "open loop control".

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# b) Feedback Control (Regelung)

We add a control device (controller) that, given an error signal e=r-y, generates the appropriate control signal u such that  $e\approx 0$ .



Question: why should an error signal (instead of r and y) be sufficient as input for the controller?

# c) Learning how to Control (similar to Feedback)

- Using the observed output y, one can define a mapping from y to u based on trials.
- Process:
  - try out best u (for given y and r), done offline
  - implement as a lookup table
- This approach might be needed if system dynamics is too fast for actuating.
- Problem: cannot be subject to quantitative analysis.

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# Open Loop Example

Classical example: cruise control

- Car driver sets car speed manually, car keeps engine power steady
- y actual speed
  - r desired speed
  - u angle of throttle (Drosselklappe)

Can be tuned for flat road, but will be off in the mountains.

### Discussion of Open Loop Control

Simple: Compute control input without continuous variable measurement

- Need to know everything accurately to work right
   Cruise-control car: friction(t), ramp\_angle(t), wind speed
- Open-loop control fails when
  - We don't know everything
  - We make errors in estimation/modeling
  - Things change

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# Feedback Control Theory

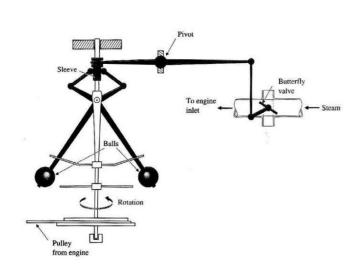
How to design the "optimal controller"?

No single solution, single approach: control theories.

- Linear vs. non-linear (differential equations)
- Deterministic vs. Stochastic
- Time-invariant vs. Time-varying control (are coefficients in controller functions of time)?
- Continuous-time vs. Discrete-time

Even: mechanical controllers!

#### Historical controllers

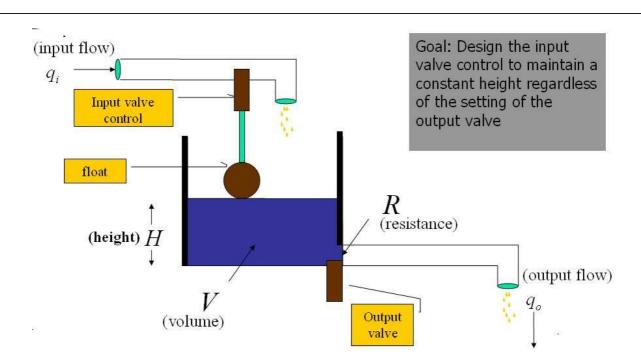


- Breeding automaton (Drebbel, ca 1620) controls heating fire based on mercury thermometer
- "Governor" for steam engine (Watts, ca 1790)
   shown to the left

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#### Another mechanical controler



# (Algorithmic) Controler Design: Basic Idea

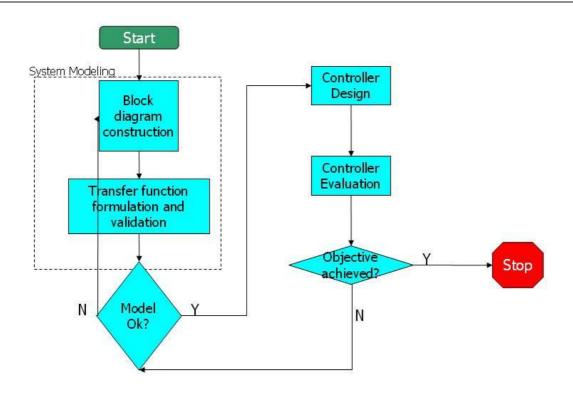
- How does (uncontrolled) plant react?
  - → capture behavior as a function (model)
- Design a (feedback) controller, to be added (controller have known dynamics, also a function)
- Goal: predict overall system (=combination of model and control function) behavior, including stablity etc

Problem: No closed theory how to find "optimal" controller.

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# Controler Design: Iterative Methodology



# **Mathematical Modelling**

Introduce "system state" x(t)

- System state x is internal to the plant, time dependent
- $\bullet$  x can be influenced (i.e., depends on) manipulated variable u
- A future system state is a function of its past state:

$$\dot{x} = f_1(x, u, t)$$

• The observable state also such a function:

$$y = f_2(x, u, t)$$

Note that x is usually not accessible, at all!

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# Mathematical Modelling (contd)

Most interesting is how the system changes:

$$\frac{d}{dt}x(t) = a(x,t) + b(x,u,t)$$

and how this affects the controlled variable (observable):

$$y(t) = c(x(t), u(t), t)$$

Usual assumption: linear behavior, time-invariant coefficients:

$$\frac{d}{dt}x(t) = Ax(t) + Bu(t)$$
$$y(t) = Cx(t) + Du(t)$$

(A...D are matrices)

Systems can, sometimes, be described without reference to internal state x, for example:  $\dot{y} + ky = u$