# Introduction

Course HECS3: Performance and quantitative properties

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### Introduction

Embedded and Cyber-Physical Systems Key Features of CPS Fundamentals of Dynamical Systems Specifying and Analyzing Properties Model-Based Design

### concepts from Embedded and Cyber-Physical Systems

- standard terminology (and some buzzwords)
- informal presentation (formalization in future lectures)
- a rough map of the territory
- what it is all for...

original computer: standalone device

**embedded system**: integrated with non-computational hardware for a specific purpose

• watches, cameras, refrigerators (integrated microcontroller), ...

more examples?

cyber-physical system<sup>1</sup>: collection of communicating computers, interacting with the physical world via feedback

- using control, computing, communication
- smart buildings, medical devices, cars, ...

example: team of autonomous robots retrieving target inside house

more examples?

<sup>&</sup>lt;sup>1</sup> term coined by Helen Gill at the US National Science Foundation (NSF) in 2006

# CPS at Verimag's Hybrid Systems Group



### Introduction

Embedded and Cyber-Physical Systems

## Key Features of CPS

- Fundamentals of Dynamical Systems
- Specifying and Analyzing Properties
- Model-Based Design

#### reactiveness

- traditionally: input  $\rightarrow$  computing  $\rightarrow$  output  $\rightarrow$  stop
- mathematically: function: inputs  $\rightarrow$  outputs
- reactive: ongoing computation
- mathematically: function from sequence of inputs to sequence of outputs

#### concurrency

- traditionally: sequential computation (Turing machine)
- concurrent: multiple threads of computation, exchanging information
- synchronous computation: all components execute in lock-step
- asynchronous computation: components act independently, communicating via messages
- both can be useful levels of abstraction

# Key Features of CPS

### feedback control

- control system interacts with physical world with sensors and actuators
- design requires modeling the dynamics of physical quantities
- traditionally: continuous dynamics

 a small enough change in the input generates a small change in the output

### real-time

- traditionally: no explicit notion of real time
- CPS: computation needs to finish within a given time frame
- timing delays, timing-dependent coordination protocols, resources allocation  $\rightarrow$  study of real-time systems

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goal: a unified view of seemingly disparate systems

- using the same concepts
- adapting techniques where necessary
- combining different techniques when systems have
  heterogeneous components

... which they do in cyber-physical systems! examples?

# Fundamentals of Dynamical Systems

### dynamical system

- precisely identified entity (we know what is part of the system and what isn't)
- defining behaviors over some notion of time (we know what "before" and "after" mean)
- with (possibly) observable outputs
- (possibly) influenced by a given set of inputs

examples for what is *not* a dynamical system?

### behavior

- evolution of states over time
- (possibly) decorated with input or output

formalized as *executions, runs, words, traces, trajectories,...* examples?

### disturbance

something that modifies the inputs or outputs of the system

random changes in the environment, electromagnetic interference, sensor noise, quantization error(!)

more examples?

### deterministic system

• if the inputs are known, there is only one future behavior

### nondeterministic system

• if the inputs are known, there is a known set of future behaviors

(actual behavior may be different each time we run the system, but belong to the same set)

### stochastic system

 if the inputs are known, the future behavior is known with a certain probability (it's the same behavior xyz% of the times we run the system)

# Fundamentals of Dynamical Systems

### state

• set of values that suffices to predict the (sets of) future behavior of the system if the inputs are known

### state-space

• the set of states of the system

example: motion of a car (with accelerator and brake pedals)

# Fundamentals of Dynamical Systems

### transition

- relates a state to a successor state
- may depend on time and inputs

### transition relation

- defines for each state the possible successor states
- a subset of states × time × inputs × states

state  $\xrightarrow{\text{time,input}}$  state'

### reachable states

- states in the closure of the transition relation
- starting from a given set of initial states

### finite state system

• the state space and the inputs are finite sets

What is maximum size of the transition relation (deterministic/nondeterministic)?

### state-space exploration (enumerative)

- starting from a given initial state, visit all reachable states, trying out all possible inputs
- = graph traversal, e.g., breadth-first search

always terminates if the state space is finite

example: check if the system can go to a given "bad" state

### infinite state system

• the state space is an infinite set (enumerable or not)

state-space exploration no longer terminates

### symbolic state-space exploration

- like state-space exploration, but using sets of states
- terminate if successors ⊆ visited states or bad states overlap visited states
- often uses overapproximation to operate on sets with simple descriptions (intervals...)

may terminate even if state space is infinite

Example: A program computing  $\sqrt{x_0}$  using the babylonian method

$$X_{k+1} = \frac{1}{2} \left( X_k + \frac{X_0}{X_k} \right).$$

implemented using int,float,rationals,reals,...

- state-space? initial state? inputs? outputs? time?
- transition relation? behaviors?
- deterministic? finite?

# Fundamentals of Dynamical Systems

Exercises:

Given an implementation using int,float,rationals,reals,...

- 1. When is enumerative state-space exploration applicable?
- 2. What are the "bad states" for checking if the sequence converges to  $\sqrt{x_0}$ ?
- 3. Apply symbolic state-space exploration starting from  $x_0 = 8$ . Use integer intervals to describe sets of states. Overapproximate if necessary.
- 4. Start from  $x_0 = 9$ . How can the precision be increased?
- 5. Does always rounding up or always rounding down cover all possibilities?

### Discrete-Time Dynamical System:

$$x_{k+1}=f(x_k,U_k).$$

- state-space? initial state? inputs? outputs? time?
- why "discrete-time"?
- transition relation?
- deterministic? finite?

Examples: Finite state machine (digital computer)

# Fundamentals of Dynamical Systems

Discrete-Time Continuous Dynamical System:

$$x_{k+1}=f(x_k,U_k).$$

- f is a continuous function of x and u:
- a small enough change in the input (or in time) generates an arbitrarily small change in the output

Examples: Digital controller (considering floating point as real numbers); sun position at noon every day

Discrete-Time Continuous Dynamical System:

$$x_{k+1}=f(x_k,U_k).$$

two main categories:

- *f* is linear:  $x_{k+1} = Ax_k + Bu_k$ either converging, diverging, or periodic
- *f* is **nonlinear**:

possibly chaotic behavior

scalar case:

$$X_{k+1} = a X_k$$

for which values of *a*:

- converging,
- diverging,
- periodic?

demographic model with reproduction and starvation [R. May, 1976]

$$x_{k+1} = rx_k(1-x_k)$$



 $x_0 = 0.6, r = 4$ 

## Fundamentals of Dynamical Systems

Discrete-Time Piecewise Continuous Dynamical System:

$$x_{k+1} = \begin{cases} f_1(x_k, u_k), & x_k \le C_1 \\ \vdots \\ f_i(x_k, u_k), & C_{i-1} < x_k \le C_i \\ \vdots \\ f_m(x_k, u_k), & x_k > C_m \end{cases}$$

• may exhibit complex behavior even for simple  $f_i$ 

Example: continuous systems with saturation of signals

### Tent Map

$$x_{k+1} = \begin{cases} \mu x_k, & x_k < \frac{1}{2}, \\ \mu (1 - x_k), & x_k \ge \frac{1}{2} \end{cases}$$



### Tent Map

$$x_{k+1} = \begin{cases} \mu x_k, & x_k < \frac{1}{2}, \\ \mu (1 - x_k), & x_k \ge \frac{1}{2} \end{cases}$$



 $x_0 = 0.6001$ ,  $\mu = 2$ 

### Tent Map



## Tent Map vs Logistic Map

tent map:

$$x_{k+1} = \begin{cases} \mu x_k, & x_k < \frac{1}{2}, \\ \mu (1 - x_k), & x_k \ge \frac{1}{2} \end{cases}$$

logistic map:

$$y_{k+1} = r y_k (1 - y_k)$$

for  $\mu = 2$  and r = 4:

$$x_k = \frac{2}{\pi} \sin^{-1} \sqrt{y_k}$$

relation between *piecewise linear* and *nonlinear* system

Continuous-Time (Continuous) Dynamical System:

Typically given by a differential equation system:

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

• can be converted to discrete-time system by sampling at time points, e.g.,  $t = k\delta$ 

Example: Motion of a car

### Discrete-Continuous Dynamical System (Hybrid System):

- mix of discrete and continuous dynamics
- discrete state changes are considered instantaneous
- discrete state determines continuous dynamics

Example: Motion of a car with gear shift

discrete (state) system (discrete dynamics) continuous system (continuous dynamics)

• discrete or continuous time

discrete-continuous (hybrid) system

What is the "right" model?

- robot turning at exactly 90 degrees, timing irrelevant: discrete system
   Can the robot leave the maze?
- maze door opens and closes at specific times: timed system

Can the robot leave the maze while the door is open?

robot not turning exactly 90 degrees:
 hybrid (discrete-continuous) system
 accumulation of deviations!
 Can the robot leave the maze while the door is open?

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- Specifying and Analyzing Properties
- Model-Based Design

### boolean properties

- state property: state → {true, false}
  e.g., predicate over the state variables
- behavior property: behavior → {true, false}
  e.g., all states along the behavior satisfy the property
- system property: system → {true, false}
  e.g., all behaviors from initial states satisfy property

these generalizations to behaviors over time are called temporal logics

# Specifying and Analyzing Properties

### probabilistic properties

- behavior probability: behavior  $\rightarrow$  [0, 1] e.g., probability that the behavior is taken (from given initial state)
- probabilistic property: system → [0, 1]
  e.g., probability that any behavior from the initial states satisfies the property

these generalizations to behaviors over time are called **probabilistic temporal logics** 

safety: nothing bad ever happens

analysis techniques:

- inductive invariants,
- state-space exploration (enumerative, symbolic)

liveness: something good eventually happens

analysis techniques:

- temporal logics,
- model checking (generalization of state-space exploration),
- ranking functions

probabilistic safety and liveness: nothing bad/something good happens with a certain probability

analysis techniques:

- probabilistic temporal logics,
- model checking,
- fault-tree analysis

**quantitative semantics** of safety and liveness: distance to nothing bad/something good happening

• e.g., min distance of any behavior to violating the property

Example:  $x(t) \le c$  for all  $t \ge 0$  (boolean safety)

• quantitative semantics  $q = \min_{t \ge 0} c - x(t)$ property satisfied iff  $q \ge 0$ .

measure of robustness<sup>2</sup>

<sup>&</sup>lt;sup>2</sup> A. Donzé, T. Ferrere, and O. Maler, "Efficient robust monitoring for STL,", in *Computer Aided Verification*, Springer, 2013, pp. 264–279.

real-time scheduling: system achieves given tasks in given time frame

analysis techniques:

- model checking,
- worst-case execution time (WCET) analysis

### quantitative property:

• computing worst-case execution time

**stability**: system will remain close to its steady state if disturbances (inputs) small enough

analysis techniques:

- linear algebra,
- Lyapunov functions (continuous ranking functions)

### quantitative property:

• *stability (gain) margin*: amount that feedback can be increased while remaining stable

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Embedded and Cyber-Physical Systems Key Features of CPS Fundamentals of Dynamical Systems Specifying and Analyzing Properties Model-Based Design

- traditionally: design, implementation, testing, validation
- model-based: formal (mathematically precise) requirements, models of the system and its environment, analysis tools for checking requirements on the model
- detect design errors earlier, ensure higher reliability

- different from programming: may incorporate nondeterminism and environment behavior
- structured design: complex tasks accomplished by composing simple components (and conversely for properties)
- requirements-based design: requirements are specified up front and guide the design (choice between design alternatives) and debugging

## Model-Based Design



#### Development Vision for Systems Mixing Software, Circuits and Mechanics (Fujitsu 2006)

http://www.fujitsu.com/downloads/EDG/binary/pdf/find/24-1e/2.pdf

## Model-Based Design



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- unambiguous not open to interpretation
- mathematically precise, can be analyzed rigorously
- block diagrams, code, state machines, differential equations

check if a formal model satisfies a property using mathematical reasoning

- rigorous (sound)
- exhaustive (all behavior is covered)
- (possibly) algorithmic or with computer support (theorem prover)

drawbacks:

- generally: not scalable (or not even decidable)
- a suitable model needs to be constructed first
- typically requires expert knowledge

hard questions:

- Does the model match reality?
- Who verifies the verifier?



- Überlingen, July 1, 2002
- 21:33:03
  - Alarm from Traffic Collision Avoidance System (TCAS)



• Überlingen, July 1, 2002

• 21:33:03

 Alarm from Traffic Collision Avoidance System (TCAS)

• 21:34:49

 Human air traffic controller command



• Überlingen, July 1, 2002

• 21:33:03

 Alarm from Traffic Collision Avoidance System (TCAS)

• 21:34:49

- Human air traffic controller command
- 21:34:56
  - TCAS recommendation



• Überlingen, July 1, 2002

• 21:33:03

 Alarm from Traffic Collision Avoidance System (TCAS)

• 21:34:49

- Human air traffic controller command
- 21:34:56
  - TCAS recommendation

• 21:35:32

- Collision





### **Formal Verification**



# Join Manoeuvre [Tomlin et al.]



- Traffic Coordination Problem
  - join paths at different speed

#### Goals

- avoid collision
- join with sufficient separation

## Join Manoeuvre



- Traffic Coordination Problem
  - join paths at different speed

#### Goals

- avoid collision
- join with sufficient separation

#### Models

- Environment: Planes
- Software: Controller
  - switches fast/slow

### Specification

keep min. distance

## Join Manoeuvre



# Join Manoeuvre



"systems that users can bet their life on"

-D. Corman, NSF

cars... airplanes... pacemakers...