Fundamentals of Dynamical Systems

Course "Feedback Control and Real-Time Systems"

Master of Science in Informatics at Grenoble Univ. Grenoble Alpes

The slides do not cover all the contents of the Feedback Control part. Some exercises are solved on the white board.

Dynamical Systems

Fundamentals of Dynamical Systems

Specifying and Analyzing Properties

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?

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)

state

 a set of independent physical quantities, the specification of which **completely** determines the 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)

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?

Exercises:

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

- 1. When is enumerative state-space exploration applicable?
- 2. How to check 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:

$$x_{k+1} = f\left(x_k, u_k\right).$$

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

Examples: Finite state machine (digital computer)

Discrete-Time **Continuous** Dynamical System:

$$x_{k+1} = f\left(x_k, u_k\right).$$

- 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:

$$x_{k+1} = f\left(x_k, u_k\right).$$

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:

$$x_{k+1} = ax_k$$

for which values of *a*:

- converging,
- diverging,
- periodic?

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



 $x_0 = 0.6, r = 4$



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



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

Tent Map

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

 $x_0 = 0.6, \, \mu = 1.5$

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:

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

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:

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

• 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?

Example: Robot in a Maze

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?

Dynamical Systems

Fundamentals of Dynamical Systems

Specifying and Analyzing Properties

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**

probabilistic properties

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

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