Cyber-Physical Systems

Laura Nenzi

Università degli Studi di Trieste Il Semestre 2022

Lecture: Examples

Artificial Pancreas

Type 1 diabetes occurs when the pancreas produces little or none of the insulin needed to regulate blood glucose

They rely on external ad- ministration of insulin to manage their blood glucose levels.



Artificial Pancreas



Stochastic Hybrid Systems Of Glucose

$$\frac{d}{dt}\boldsymbol{x}(t) = F(\boldsymbol{x}(t); u(t); \boldsymbol{\Theta});$$

$$y(t) = x_1(t)$$

Time (min)

glucose concentration

Infusion rate of bolus insulin the control parameters

 $\Theta = (D_{G_1}; D_{G_2}; D_{G_3}; T_1; T_2)$ are the control parameter

 $(D_{G_1}; D_{G_2}; D_{G_3}) \in (N(40; 10); N(90; 10); N(60; 10))$ are the three daily meals

 $(T_1; T_2) \in \sim N$ (300, 10) and $T_2 \sim N$ (300, 10) are the inter-times between each of them



Stochastic Hybrid Systems Of Glucose

$$\frac{d}{dt}Q_{1}(t) = -F_{01} - x_{1}Q_{1} + k_{12}Q_{2} - F_{R} + EGP_{0}(1 - x_{3}) + \frac{D_{G}A_{G}}{t_{maxG}^{2}}te^{-\frac{t}{t_{maxG}}}$$

$$\frac{d}{dt}Q_{2}(t) = x_{1}Q_{1} - (k_{12} + x_{2})Q_{2};$$

$$\frac{d}{dt}S_{1}(t) = u(t) + u_{b} - \frac{S_{1}}{t_{maxI}};$$

$$\frac{d}{dt}S_{2}(t) = S_{1} - \frac{S_{2}}{t_{maxI}};$$

$$\frac{d}{dt}I(t) = \frac{S_{2}}{t_{maxI}V_{I}} - keI;$$

$$\frac{d}{dt}x_{i}(t) = -k_{a_{i}}x_{i} + k_{b_{i}}I; \quad (i = 1, 2, 3)$$

$$u(t) = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \frac{d}{dt} e(t), \qquad e(t) = r(t) - y(t)$$

PID Control



$$u(t) = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \frac{d}{dt} e(t), \qquad e(t) = r(t) - y(t)$$

Artificial Pancreas Simulation



STL Properties for the Artificial Pancreas

►<u>Hyperglycemia</u>

"during the day the level of glucose goes above 180mg/dl"

 $\neg G_{[0,24h]}(BG(t) < 180)$



► <u>Hypoglycemia</u>

"during the day the level of glucose goes below 70mg/dl"

 $\neg G_{[0,24h]}(BG(t) > 70)$

STL Properties for the Artificial Pancreas

Prolonged Hyperglycemia

"during the day the level of glucose goes above 180mg/dl for 3 hours"

 $F_{[0,21h]}(G_{[0,3]}(BG(t) \ge 180))$

Prolonged Hypoglycemia

"during the day the level of glucose goes below 70mg/dl for 30 minutes"

 $F_{[0,21h]}(G_{[0,0.5]}(BG(t)<70)$



Falsification

The most simple way to do falsification with respect a property ϕ is minimizing the robustness over N iterations considering random samples on control parameters, i.e.

```
\begin{split} \min STL &= \inf' \\ \text{For } i = 1, \dots, \text{N}: \\ \Theta &= \text{sampling } (D_{G_1}, D_{G_2}, D_{G_3}, T_1, T_2) \\ \text{t}, y &= \text{simulation}(\Theta) \\ \text{stl} &= \text{computeRobustness}(y, \varphi) \\ \text{if } (\text{stl} < \min STL): \\ \min STL &= \text{stl} \\ \text{vSTL} &= [D_{G_1}, D_{G_2}, D_{G_3}, T_1, T_2] \end{split}
```

For fixed control parameter spaces you can consider to sample with respect to the grids over it.

Noise Robustness

► To consider noisy sensor we can add a Gaussian noise to the generated glucose trajectory, i.e. $GB(t) + \gamma$ with $\gamma \in N(0; 5)$



Automatic Transmission



Most material that follows is from this paper:

Bardh Hoxha, Houssam Abbas, Georgios E. Fainekos: Benchmarks for Temporal Logic Requirements for Automotive Systems. ARCH@CPSWeek 2014: 25-30

Automatic Transmission

- Inputs: the throttle and break
- Outputs: the speed of the engine ω (RPM), the speed of the vehicle v (mph) and the gear.
- Initially, the vehicle is at rest at time 0, i.e. the speed v = 0 and engine speed ω = 0
- Therefore, the output trajectories depend only on the input signals ut and ub which model the throttle and break inputs.
- The throttle and break, at each point in time, can take any value between 0 (fully closed) to 100 (fully open).





Automatic Transmission

The model contains 69 blocks among which there are 2 integrators (i.e., 2 continuous state variables), and a Stateflow chart. The Stateflow chart contains two concurrently executing Finite State Machines with 4 and 3 states, respectively.





Properties

Automatic Transmission				
	Natural Language	MTL		
ϕ_1^{AT}	The engine speed never reaches $\bar{\omega}$.	$\Box(\omega<\bar{\omega})$		
ϕ_2^{AT}	The engine and the vehicle speed never reach $\bar{\omega}$ and \bar{v} , resp.	$\Box((\omega<\bar{\omega})\wedge(v<\bar{v}))$		
ϕ_3^{AT}	There should be no transition from gear two to gear one and back to gear two in less than 2.5 sec.	$\Box((g_2 \wedge Xg_1) \to \Box_{(0,2.5]} \neg g_2)$		
ϕ_4^{AT}	After shifting into gear one, there should be no shift from gear one to any other gear within 2.5 sec.	$\Box((\neg g_1 \land Xg_1) \to \Box_{(0,2.5]}g_1)$		
ϕ_5^{AT}	When shifting into any gear, there should be no shift from that gear to any other gear within 2.5sec.	$\wedge_{i=1}^4 \Box((\neg g_i \wedge Xg_i) \to \Box_{(0,2.5]}g_i)$		
ϕ_6^{AT}	If engine speed is always less than $\bar{\omega}$, then vehicle speed can not exceed \bar{v} in less than T sec.	$\neg(\diamondsuit_{[0,T]}(v > \bar{v}) \land \Box(\omega < \bar{\omega}))$		
ϕ_7^{AT}	Within T sec the vehicle speed is above \bar{v} and from that point on the engine speed is always less than $\bar{\omega}$.	$\diamondsuit_{[0,T]}((v \ge \bar{v}) \land \Box(\omega < \bar{\omega}))$		
ϕ_8^{AT}	A gear increase from first to fourth in under 10secs, ending in an RPM above $\bar{\omega}$ within 2 seconds of that, should result in a vehicle speed above \bar{v} .	$((g_1 \ \mathcal{U} \ g_2 \ \mathcal{U} \ g_3 \ \mathcal{U} \ g_4) \land \diamondsuit_{[0,10]}(g_4 \land \diamondsuit_{[0,2]}(\omega \ge \bar{\omega}))) \to \diamondsuit_{[0,10]}(g_4 \to X(g_4 \ \mathcal{U}_{[0,1]} \ (v \ge \bar{v})))$		



Pacemaker



Electrical Conduction System of the Heart

- SA node (controlled by nervous system) periodically generates an electric pulse
- This pulse causes both atria to contract pushing blood into the ventricles
- Conduction is delayed at the AV node allowing ventricles to full fill
 - Finally the His-Purkinje system spreads electric activation through ventricles causing them both to contract, pumping blood out of the heart

What do pacemakers do?



- Aging and/or diseases cause conduction properties of heart tissue to change leading to timing anomalies in heart rhythm (arrhythmias)
- Tachycardia: faster than desirable heart rate impairing hemo-dynamics (blood flow dynamics)
- Bradycardia: slower heart rate leading to insufficient blood supply
- Pacemakers can be used to treat bradycardia by providing pulses when heart rate is low

How dual-chamber pacemakers work

- Two fixed leads on wall of right atrium and ventricle respectively
- Activation of local tissue sensed by the leads (giving rise to events Atrial Sense (AS) and Ventricular Sense (VS))
- Atrial Pacing (AP) or Ventricular Pacing (VP) are delivered if no sensed events occur within deadlines



Implantable Pacemaker modeling



The LRI mode of operation explained



K= 850ms

- LRI (Low Rate Interval) component keeps heart rate above minimum level
- One of the pacemaker modes of operation that models the basic timing cycle
- Measures the longest interval between ventricular events
- Clock reset when VS or VP received
- No AS received ⇒ LRI outputs AP after K (TLRI-TAVI) time units



Hodgkin - Huxley model

The whole heart model consists of a linear composition of cell models, which synchronize according to their output and input stimuli

At the top of the network, we have the sinoatrial (SA) node: it's input stimulus can come from the natural pacing of the heart or from pacemaker's actuator.



Random Heart Model (RHM)

RHM is designed to cover open-loop heart behaviors For the atrial region for instance, the interval between each action (*Aget!*) is a random value from the interval (*Aminwait, Amaxwait*).



Property

TCTL formula : $A \Box ((VS! VVP!) \rightarrow A \Box_{\leq TLRI} (VS! VVP!))$

The interval between two ventricular events should be less than TLRI





TCTL formula: $A\Box(ch1[0] ! \rightarrow A\Diamond ch1[N] !)$

Given an initial input, the signal should propagate all the way from SA node to atrium and then ventricle, and eventually be visible at the end of the N cells chain described in the previous section.

Temperature Control of a Continuous Stirred Tank Reactor

- Control (PID and MPC) the temperature of an exothermic CSTR so that it follows a constant set point;
- Requirement specification and checking using STL;
- Falsification of the requirements;



Plant Model

First Order Reaction: $A \rightarrow_{k \cdot C_A} B$ Reaction rate per unit volume (Arrhenius law): $k(T) = k_0 e^{-E_a/R \cdot T}$

Mole balance equation:

$$\frac{dC_A}{dt} = \frac{q}{V} (C_{Af} - C_A) - k(T)C_A$$

Energy balance equation:

$$\frac{dT}{dt} = \frac{q}{V} \left(T_f - T \right) + \frac{-\Delta H_R}{\rho C_p} k(T) C_A + \frac{UA}{\rho C_p V} \left(T - T_c \right)$$

Input constraint: $T_C \in [250, 350]$



Simulation of the plant

Depending on the reactant A and on the product that we want to obtain, it is desirable that the internal temperature of the reactor T sets to a constant reference value (low if we want concentration of A to be high, and viceversa). However, without any form of control, this cannot be easily achieved, due to the dynamics of the system:



Simulation of the sensor

Same as before, but adding Gaussian Noise to observations



PID Control Cont'd

0

Performance of PID controller when tracking a temperature of 355K



Non-linear MPC Cont'd

0

Performance of MPC when tracking a temperature of 375K



STL Requirements 3

0

Goal: CSTR should closely follow reference temperature

Difference from reference: $d(t) = |T(t) - ref(t)| \forall t$

1. In the last part of the simulation, difference from reference should not exceed 3K



Falsification Cont'd

0

MPC parameters found in falsification analysis

Reference	\mathbf{Q}	R	Robustness
320	1.33663	0.020416	-1.28741
325	0.376885	0.024445	-20.7467
330	0.210134	0.0191757	-31.3692
335	1.60591	0.0040131	-0.423672
340	1.63118	0.00144311	-0.0405561
345	2.45878	0.0092065	-0.10704
350	2.67307	0.0045229	-2.28309
355	1.57202	0.00161315	-3.23189
360	1.67468	0.0150942	-8.42903
365	2.70502	0.00979266	-58.8853
370	2.59014	0.00121103	-5.30934
375	1.73716	0.0150608	-0.0113018
380	1.48972	0.018182	-2.94381
385	0.226472	0.0107259	-8.14433



Falsification for reference temperature 370K

A Deep Neural Network controller



Mojtaba Zarei, Yu Wang, Miroslav Pajic: <u>Statistical verification of learning-based cyber-physical</u> systems. HSCC 2020: 12:1-12:7

Bibliography

Nice survey on Specification-Based Monitoring of CPSs: http://www-verimag.imag.fr/PEOPLE/maler/Papers/monitor-RV-chapter.pdf

Artificial Pancreas:

- F. Shmarov, N. Paoletti, E. Bartocci, S. Lin, S. A. Smolka, and P. Zuliani. Automated synthesis of safe and robust PID controllers for stochastic hybrid systems. arXiv:1707.05229, 2017.
- Simone Silvetti, Laura Nenzi, Ezio Bartocci, Luca Bortolussi: Signal Convolution Logic. CoRR abs/1806.00238 (2018)
- Fraser Cameron, Georgios E. Fainekos, David M. Maahs, Sriram Sankaranarayanan: Towards a Verified Artificial Pancreas: Challenges and Solutions for Runtime Verification. RV 2015: 3-17
- Sriram Sankaranarayanan, Suhas Akshar Kumar, Faye Cameron, B. Wayne Bequette, Georgios E. Fainekos, David M. Maahs:Modelbased falsification of an artificial pancreas control system. SIGBED Rev. 14(2): 24-33 (2017)

Pacemaker:

- Z. Jiang, M. Pajic, S. Moarref, R. Alur, R. Mangharam, *Modeling and Verification of a Dual Chamber Implantable Pacemaker*, In Proceedings of Tools and Algorithms for the Construction and Analysis of Systems (TACAS), 2012.
- The textbook has detailed descriptions of some other pacemaker components