#### Is it about time for control?

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when is (the right) time for control?



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We know very little about the role of *time* in control although there are far too many misconceptions.

Let me argue as we sip different cocktails of time and control.



The context

We start with a textbook example to provide a concrete context for our cocktails.



$$\dot{x} = v$$
  
 $\dot{v} = -\frac{k}{m}x + \frac{1}{m}u$ 

The objective is to design a feedback control law u = Kx to stabilize the mass at x = 0.

For simplicity we take m = 1 and k = 10.

Existing control design approaches

Several alternatives are possible:

- 1 Periodic control (classical approach, first part of my talk);
- 2 Event-triggered, self-triggered control and beyond (second part of my talk).



Classical approach:

The first cocktail

**1** Design a controller (u = -4v) resulting in an asymptotically stable<sup>1</sup> closed-loop system assuming ideal sensors, ideal actuators, and instantaneous infinite precision computation.



<sup>1</sup> With eigenvalues at  $\lambda = -2 \pm i\sqrt{6}$ .

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- 2 Implement this controller in a periodic sample-and-hold manner.



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- 2 Implement this controller in a periodic sample-and-hold manner.

It is known that under a sufficiently fast periodic sample-and-hold implementation, the implemented closed-loop system will retain stability.

Perhaps due to this result, it is widely believed that control requires short periods (high sampling rates) and that stability is lost if deadlines are missed or periods are increased.

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The first cocktail

For a periodic implementation asymptotic stability is equivalent to the closed-loop eigenvalues being strictly inside the unit disk.

Let us plot the location of the eigenvalues as a function of the chosen period.





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Stability is achieved with small periods as well as with (arbitrarily) large periods!

The second cocktail: adding the performance ingredient

Stability, however, is not the full story.





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Is performance monotonically increasing with decreasing periods?



The second cocktail: adding the performance ingredient

Stability, however, is not the full story.



Is performance monotonically increasing with decreasing periods?

No! Shorter periods magnify the undesired effects of sensor noise and quantization.<sup>2</sup>

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<sup>&</sup>lt;sup>2</sup> B. Bamieh. Intersample and finite wordlength effects in sampled-data problems. IEEE Transactions on Automatic Control, 48(4), 639–643, 2003.

The third cocktail: mixing periods

Periods and performance, different cocktails for different tastes.





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The third cocktail: mixing periods

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 $J(1,1) \approx 25.4$ 

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UCLA

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The third cocktail: mixing periods

We can even stabilize this system using only periods for which the system is unstable!





Reflections of an inebriated man

 The relation between time, stability, and performance is quite complex and not well understood<sup>3</sup>;



 $^{3}$  Beware of good looking (convex) Quality of Control or utility functions.

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- The relation between time, stability, and performance is quite complex and not well understood<sup>3</sup>;
- The periodic paradigm only offers a narrow view of the role of time in control. In particular, it fails to explain why we obtain better performance when using unstable periods;
- The periodic paradigm results in a conservative usage of resources since the period is chosen for worst case operating conditions that rarely happen;
- The periodic paradigm can be seen as open-loop sampling;
- Recently, event-triggered and self-triggered control have been proposed as a way of introducing *feedback* in the sampling process.



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In time-triggered control the sensing, control, and actuation are driven by a clock.





In event-triggered control the input is held constant, not periodically, but while performance is satisfactory.



This requires the constant monitoring of the state to determine the current performance.

In self-triggered control, the current state is used not only to compute the input to the system, but also the next time the control law should be recomputed;



Constant monitoring of the state is no longer needed although the loop is still closed based on the current performance.



#### Event-triggered and self-triggered control A first example of self-triggered control

As an example we consider the control of a jet engine compressor:



$$\begin{aligned} \dot{x}_1 &= -x_2 - \frac{3}{2}x_1^2 - \frac{1}{2}x_1^3 \\ \dot{x}_2 &= \frac{1}{\beta^2}(x_1 - u) \\ u &= x_1 - \frac{\beta^2}{2}(x_1^2 + 1)(y + x_1^2y + x_1y^2) + 2\beta^2 x_1 \\ y &= 2\frac{x_1^2 + x_2}{x_1^2 + 1} \end{aligned}$$

New deadline is a function of the current state of the physical system:

$$\tau(x_1, y) = \frac{29x_1 + x_1^2 + y^2}{5.36x_1^2\sqrt{x_1^2 + y^2} + x_1^2 + y^2} \cdot \tau^*.$$

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It works in theory... How about in practice?



Welcome to the real world!

In collaboration with Kalle Johansson at KTH we tested these ideas in a real deployment.



- Two independent water tanks controlled over the same wireless network.
- 1 sensing, 1 controller, and 1 actuator node per water tank.
- Additional temperature and humidity sensors on the wall to generate low priority traffic.

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- Slight modification of the IEEE 802.15.4 MAC protocol (used in the proposed WirelessHART);
- The IEEE 802.15.4 MAC protocol offers a Contention Access Period (CAP) using CSMA/CA and a Contention Free Period (CFP) using TDMA;
- Telos motes with 250kbps 2.4 GHz Chipcon CC2420 IEEE 802.15.4 compliant radio.

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Scheme	Integral of Absolute Error	#updates in 220 s.	Battery life (days)
Event-triggered	88.26	49	879.84
Self-triggered	101.65	34	1010.18
Periodic	88.50	347	636.81



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- Fully exploiting the promise of these new paradigms requires more research to understand:
  - how to dynamically trade computation and communication resources for performance;
  - how to mix time-triggered with event-triggered controllers;
  - new generation of:
    - sensors/actuators;
    - real-time scheduling algorithms;
    - communication protocols,

that natively support mixed time-triggered and event-triggered algorithms;



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For papers and more information:

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http://www.cyphylab.ee.ucla.edu
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