

Advanced Topics in Control: Distributed Systems and Control

Lecture 9: Introduction to Event-Triggered Control

Dr. Mohammad H. (Vahid) Mamduhi, April 29th, 2022



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Summary of Lecture 8 on Stochastic Stability

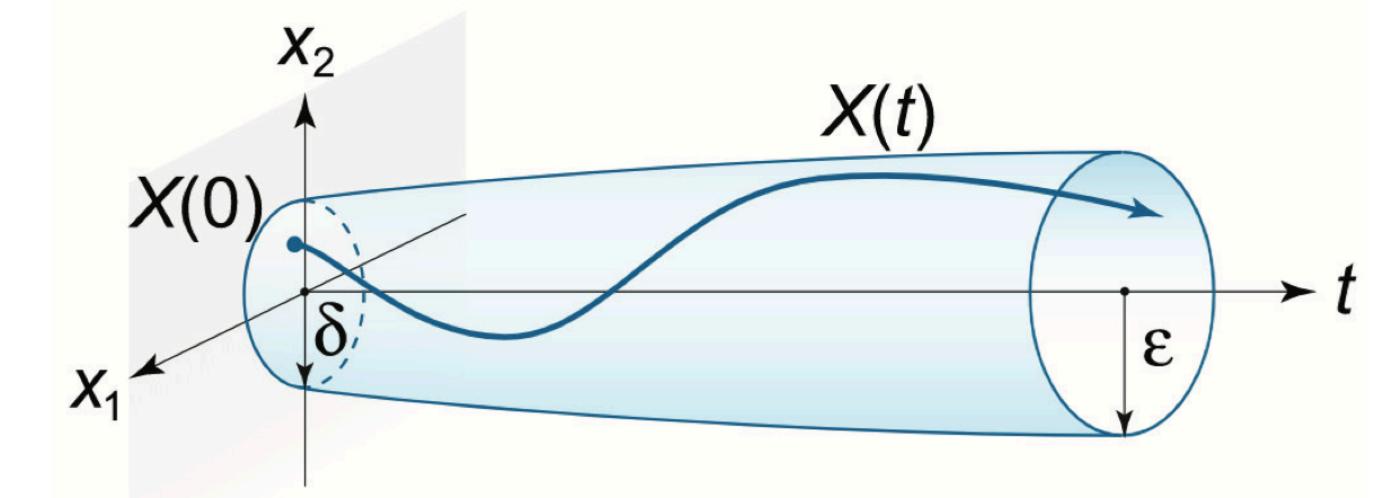
Consider the discrete-time stochastic dynamical system

$$x(t+1) = f(x(t), w(t)), \quad x(t_0) = x(0)$$

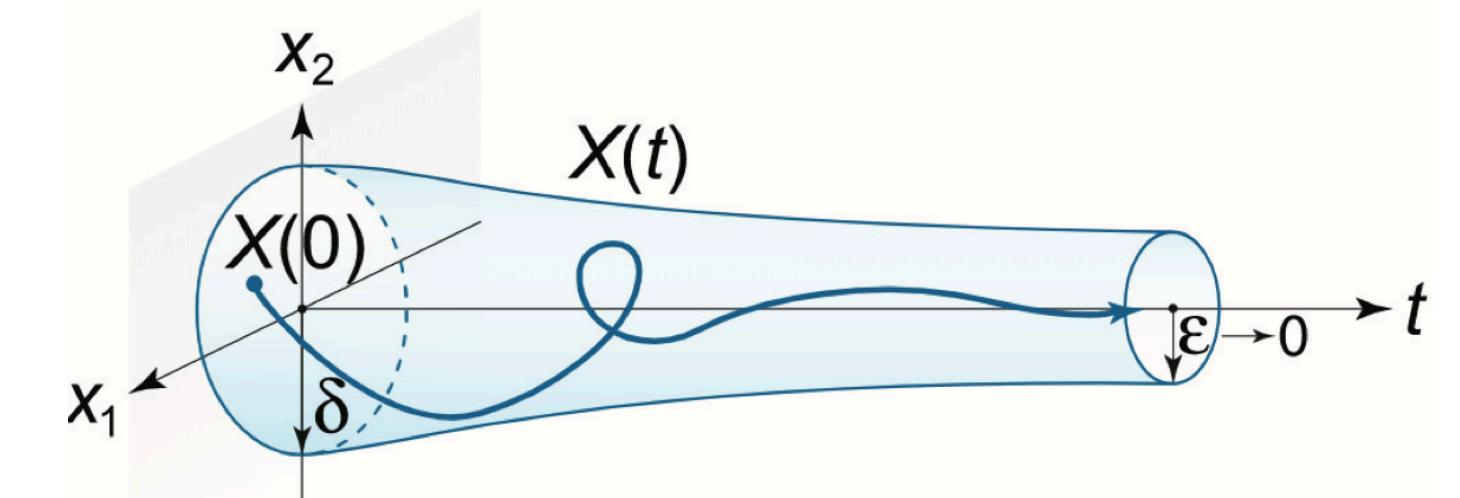
f is a (Lipschitz) continuous function and $w(t), t \in \mathbb{N}_0$ is a sequence of Markov random variables. We are interested in stability of the equilibrium $x(t, x(0)) = 0$

Recall stability notions for deterministic system $x(t+1) = f(x(t))$. $\forall \varepsilon > 0, \exists \delta > 0$:

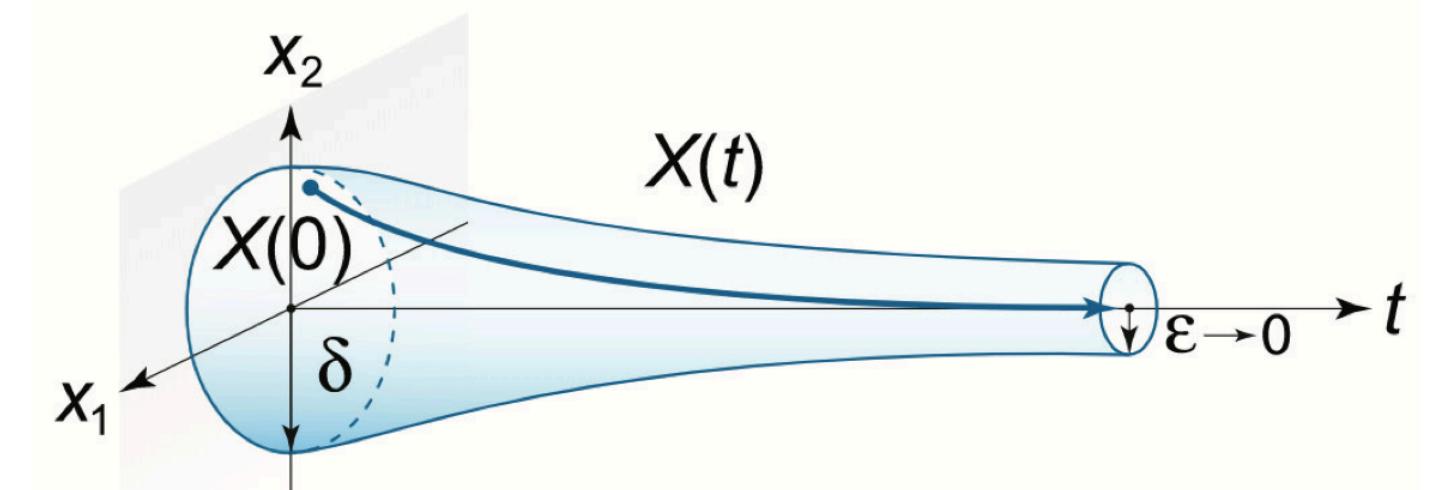
1. **Lyapunov stability:** trajectory starts in δ -vicinity of the equilibrium, always remains in its ε -vicinity
2. **Asymptotic stability:** trajectory starts in δ -vicinity of equilibrium, it stays in its ε -vicinity and eventually converges to the equilibrium
3. **Exponential stability:** trajectory starts in δ -vicinity of equilibrium, it remains in its ε -vicinity and smoothly converges to equilibrium with a particular rate



Lyapunov stability



Asymptotic stability



Exponential stability

Summary of Lecture 8 on Stochastic Stability

Exponential stability \Rightarrow Asymptotic stability \Rightarrow Lyapunov stability

Generally for stochastic systems, stability is discussed in the sense of expectation or probability, because trajectories vary from one sample to another due to random dynamics, i.e., $w(t)$

$$x(t+1) = f(x(t), w(t)), \quad x(t_0) = x(0)$$

Lyapunov-based stochastic stability:

1. Lyapunov stability in probability: For any $\varepsilon_1 > 0$, $\varepsilon_2 \in [0,1]$, there exists $\delta(\varepsilon_1, \varepsilon_2, t_0)$, s.t. $\|x(0)\| < \delta$ implies

$$\mathbb{P}\left\{\sup_{t>t_0} \|x(t, x_0)\| > \varepsilon_1\right\} < \varepsilon_2$$

2. Lyapunov stability in the m^{th} mean: For any $\varepsilon > 0$, there exists $\delta(\varepsilon, t_0)$, s.t. $\|x(0)\|_m < \delta$ implies

$$\mathbb{E}\left\{\sup_{t>t_0} \|x(t, x(0))\|_m^m\right\} \leq \varepsilon$$

Summary of Lecture 8 on Stochastic Stability

Asymptotic stochastic stability:

1. Asymptotic stability in probability: if Lyapunov stable in probability and $\exists \delta > 0$ s.t. $\|x(0)\| < \delta$ implies

$$\lim_{\tau \rightarrow \infty} \mathbb{P}\{\sup_{t \geq \tau} \|x(t, x(0))\| > \varepsilon\} = 0$$

2. Asymptotic stability in the m^{th} mean: if Lyapunov stable in m^{th} mean and $\exists \delta > 0$ s.t. $\|x(0)\| < \delta$ implies

$$\lim_{\tau \rightarrow \infty} \mathbb{E}\{\sup_{t \geq \tau} \|x(t, x(0))\|_m^m\} = 0$$

Exponential stochastic stability:

- Exponential stability of the m^{th} mean: if $\exists \delta > 0$ and constants α, β s.t. $\|x(0)\| < \delta$ implies

$$\mathbb{E}\{\|x(t, x_0)\|_m^m\} \leq \beta \|x_0\|_m^m e^{-\alpha(t-t_0)}$$

- Exponential stability of m^{th} mean $\not\Rightarrow$ Asymptotic stability in m^{th} mean \Rightarrow Lyapunov stability in m^{th} mean
- Stability in the mean is stronger than stability in probability!

Event-Triggered Scheme

When execution of an action (control, sampling, actuation, communication, etc.) is performed based on the occurrence of a **defined “event”**, we call the action scheme ***event-triggered*** or ***event-based***.

Compared with **Time-Triggered** scheme, where an action is executed at pre-defined and equidistant time instances!

We do things in event-based fashion in our daily lives as well, e.g., when sick:

- ▶ Take medication every 8 hours → Time-triggered
- ▶ Take medication If body temperature above 38 → Event-triggered



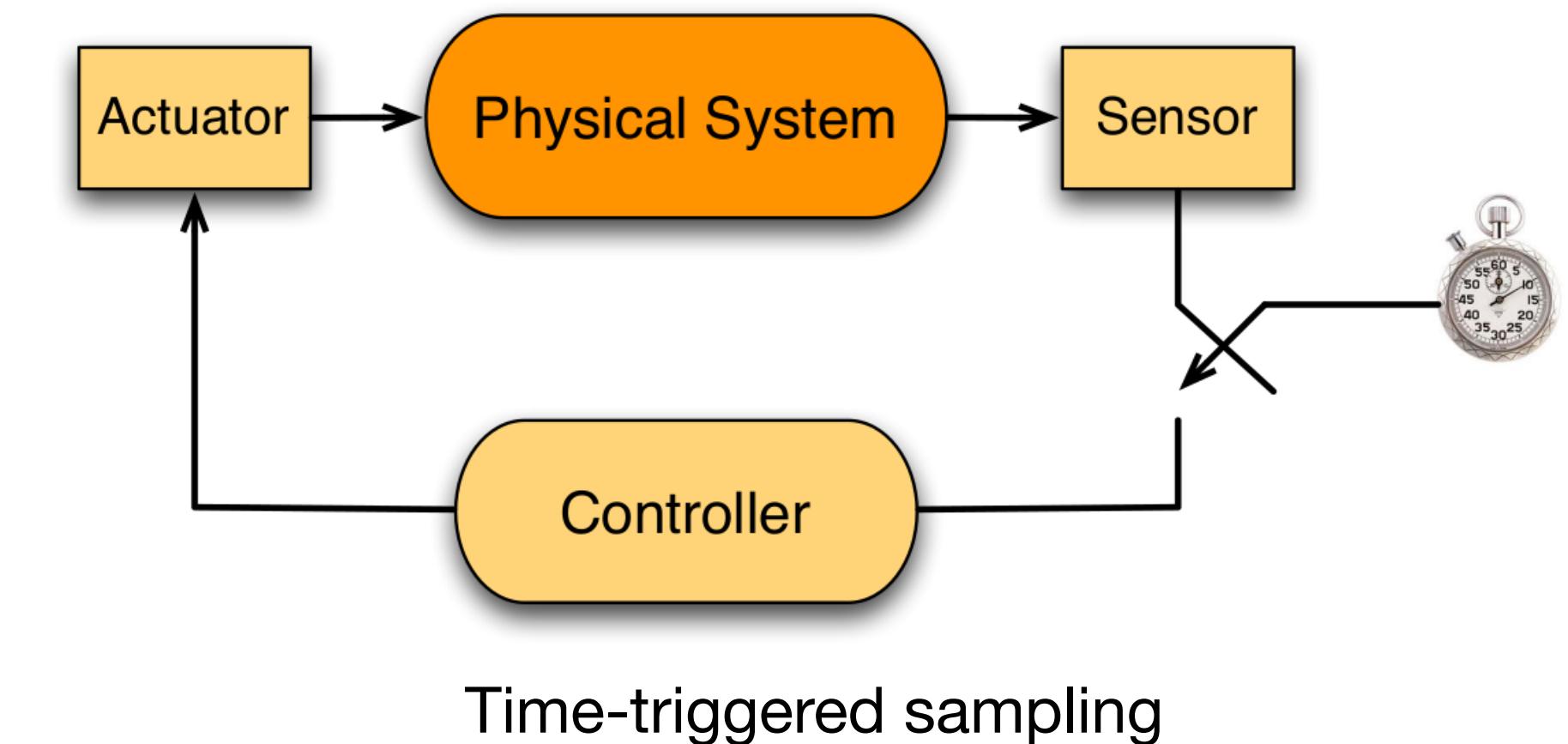
Intuition: Actions often entail costs (implementation/resources), so why not perform an action only when necessary?

Major benefit: reduction of actions and consequently reduction of consuming costly resources

Event-Triggered vs. Time-Triggered Sampling

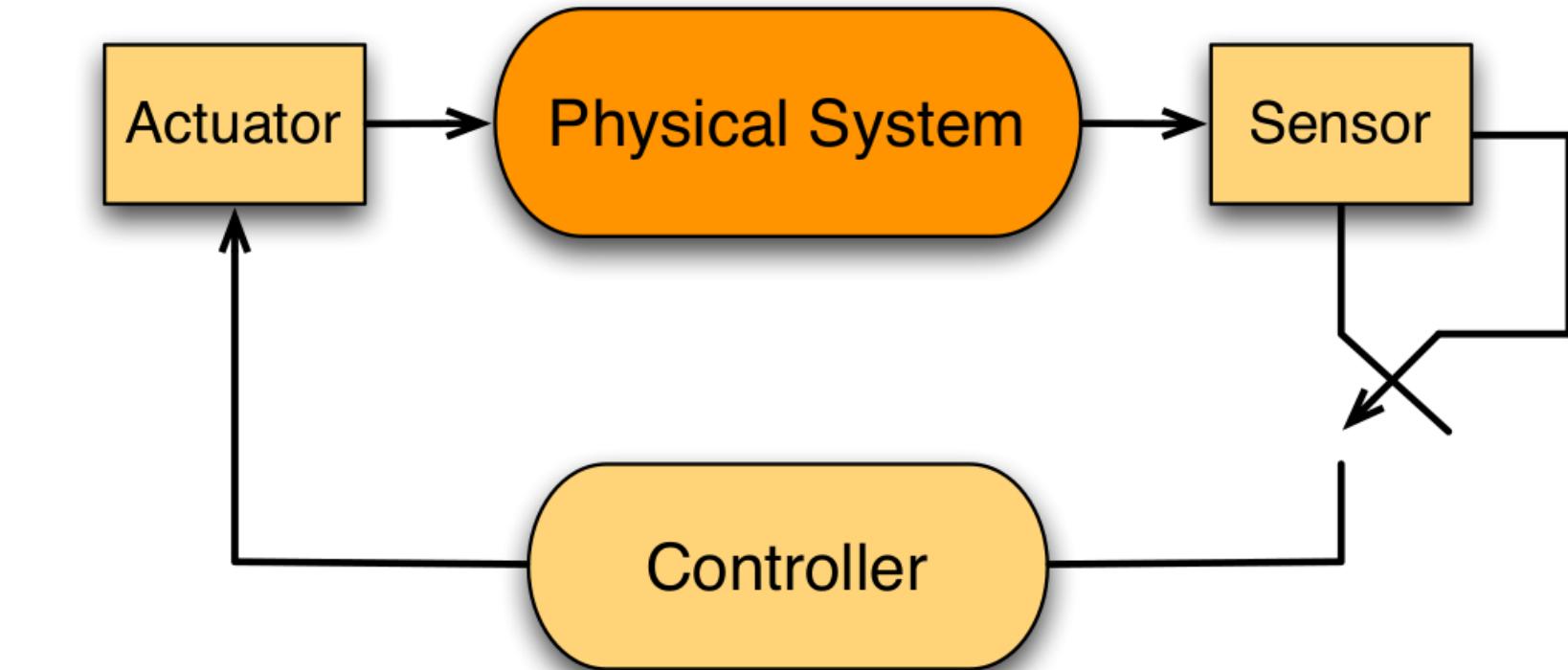
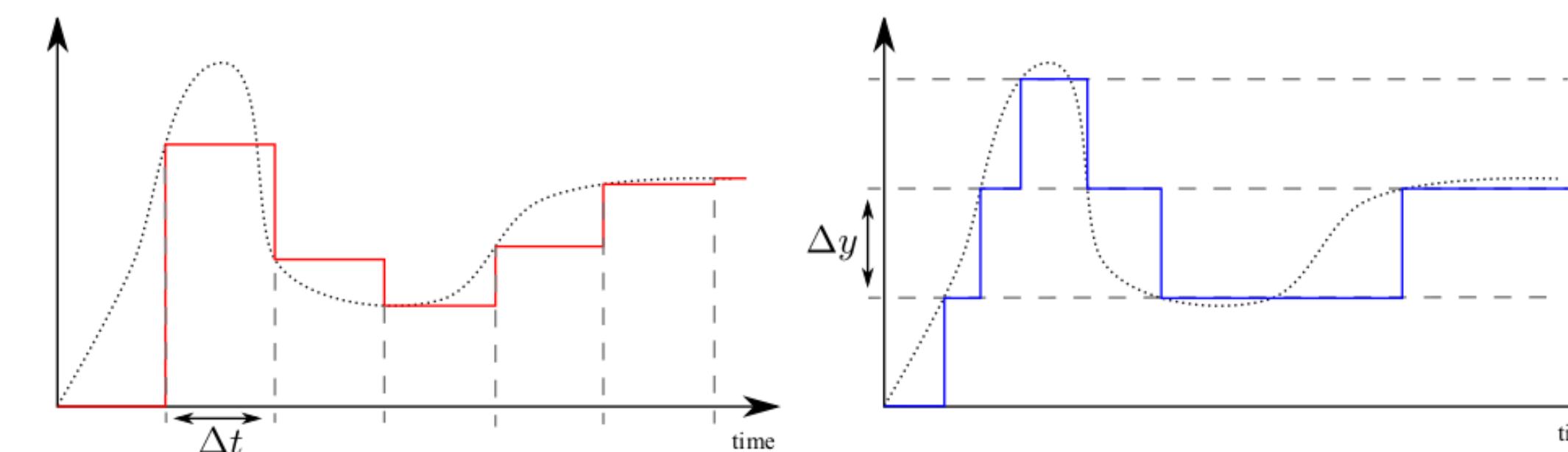
Time-triggered sampling is an open-loop scheme where sampling is performed using a clock

- Pre-specified sampling instances
- Deterministic and synchronous sampling
- Not resource efficient and not flexible



Event-triggered sampling uses feedback in the sampling process

- Real-time sampling instances
- Asynchronous sampling and difficult to predict
- Resource efficient (sample only when performance not satisfactory)



Event-Triggered vs. Time-Triggered Sampling: A more complex scenario

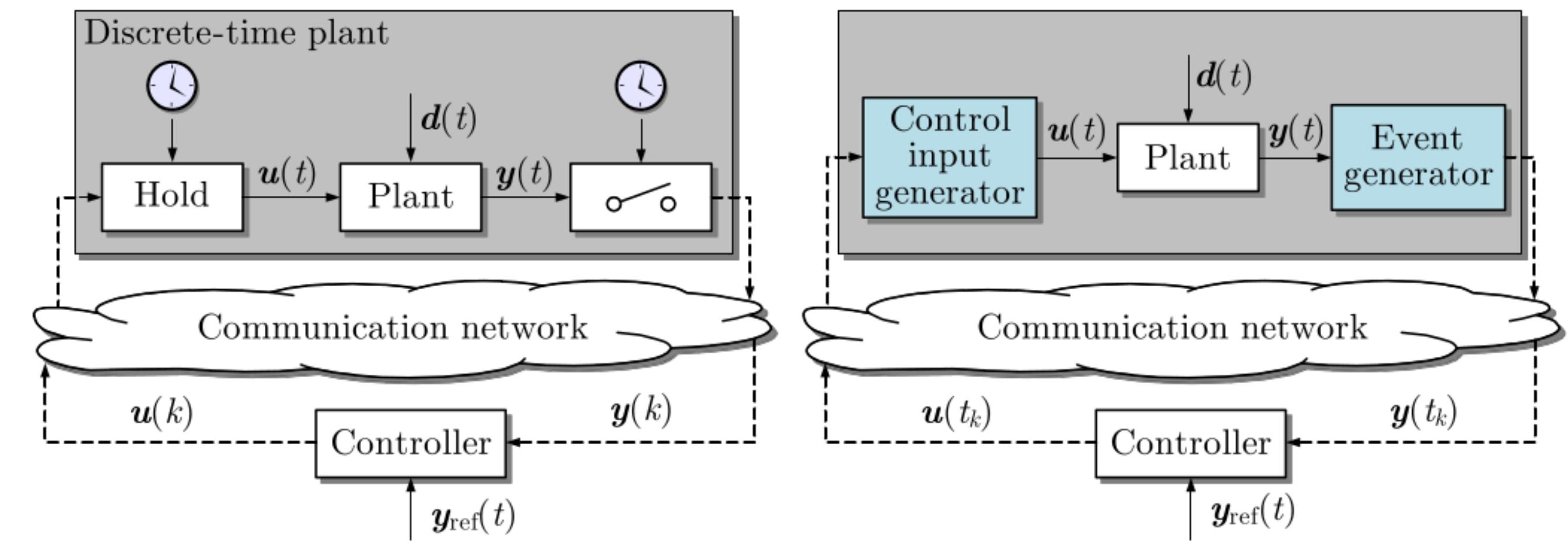
Event-triggered actions: events can be designed for various functionalities across a control system

A single control system

- When sample the system states
- When update the control input/actuate

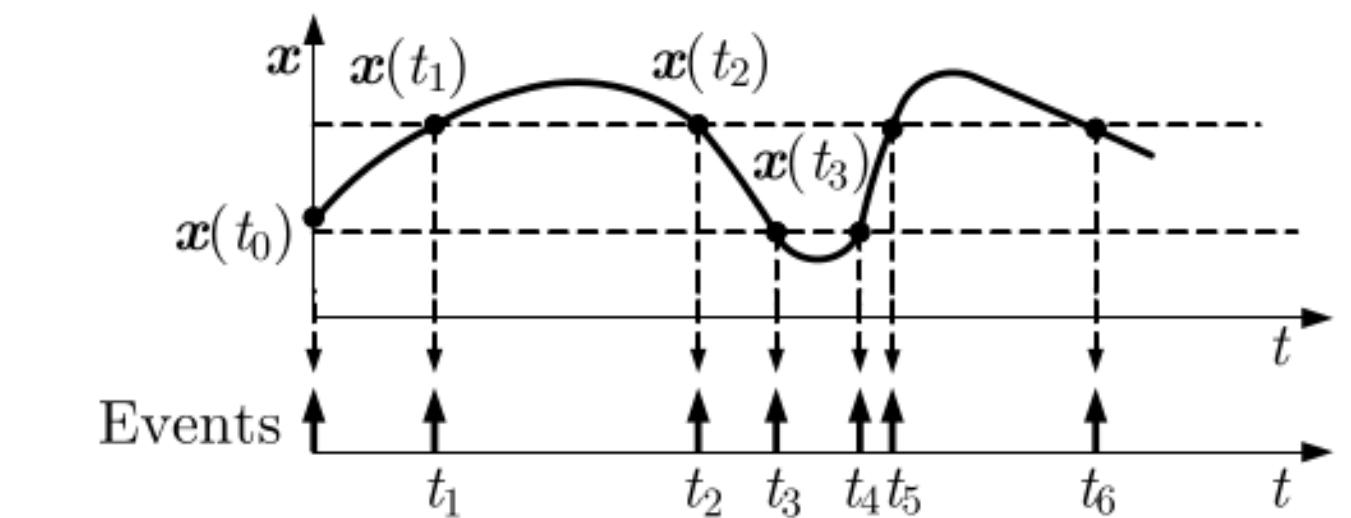
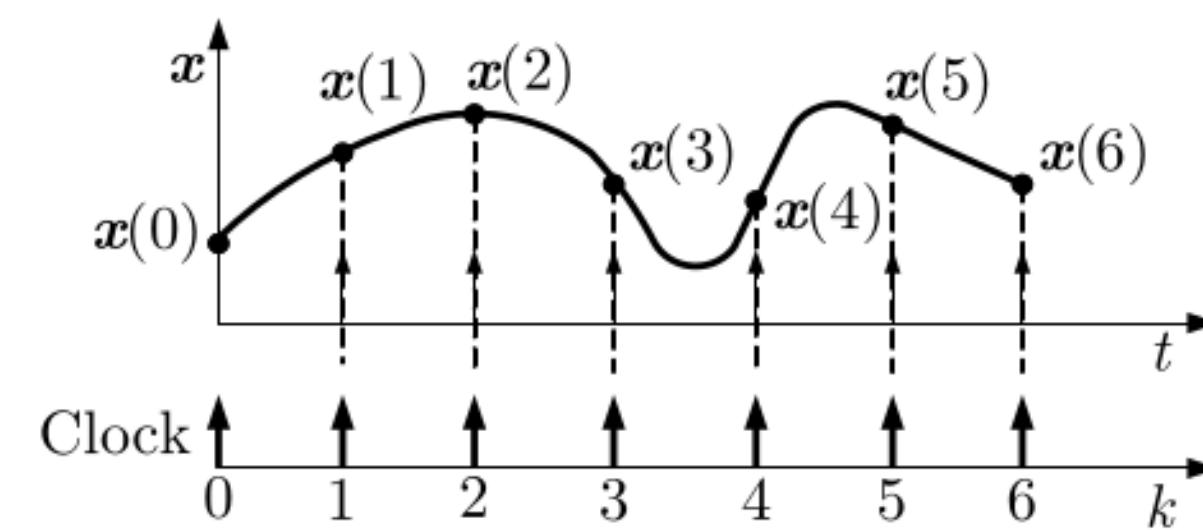
A networked/multi-agent system

- When to communicate and with whom
- Which loop to be closed, if resources limited



Challenges of event-triggered schemes:

- Complex to design and implement
- Optimal events often difficult to be found
- Requires constant monitoring of the events
- Triggering instances often not predictable

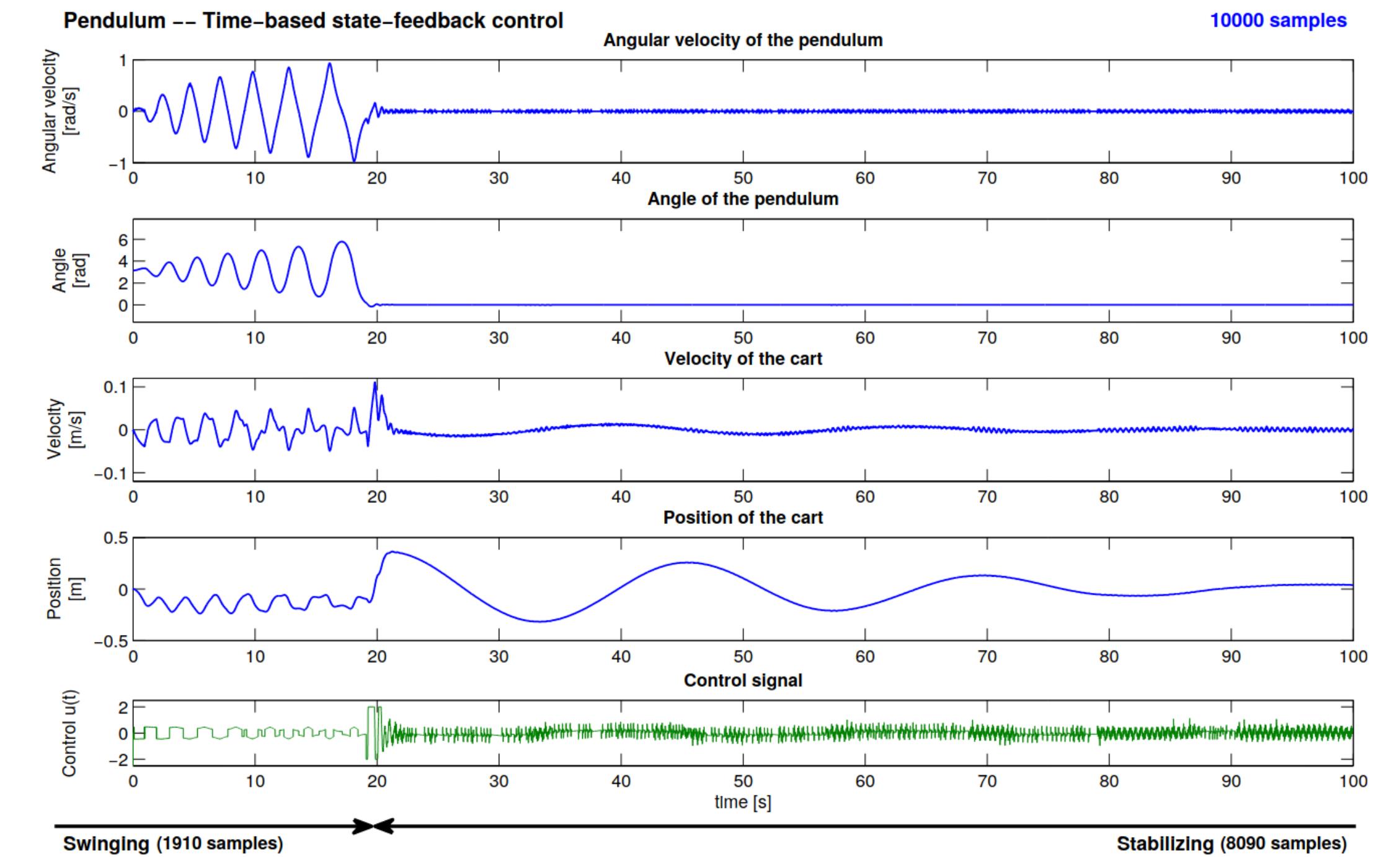
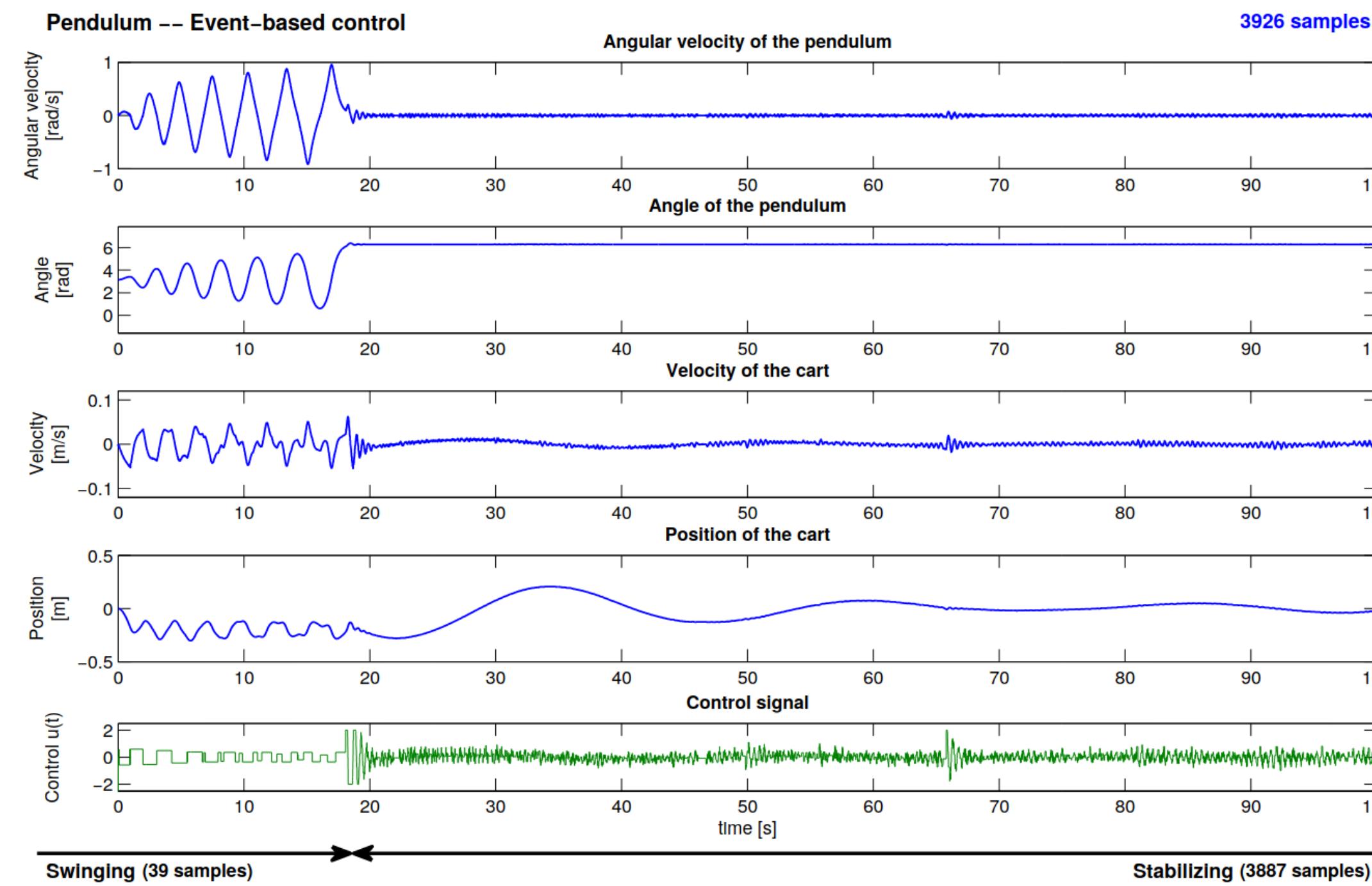
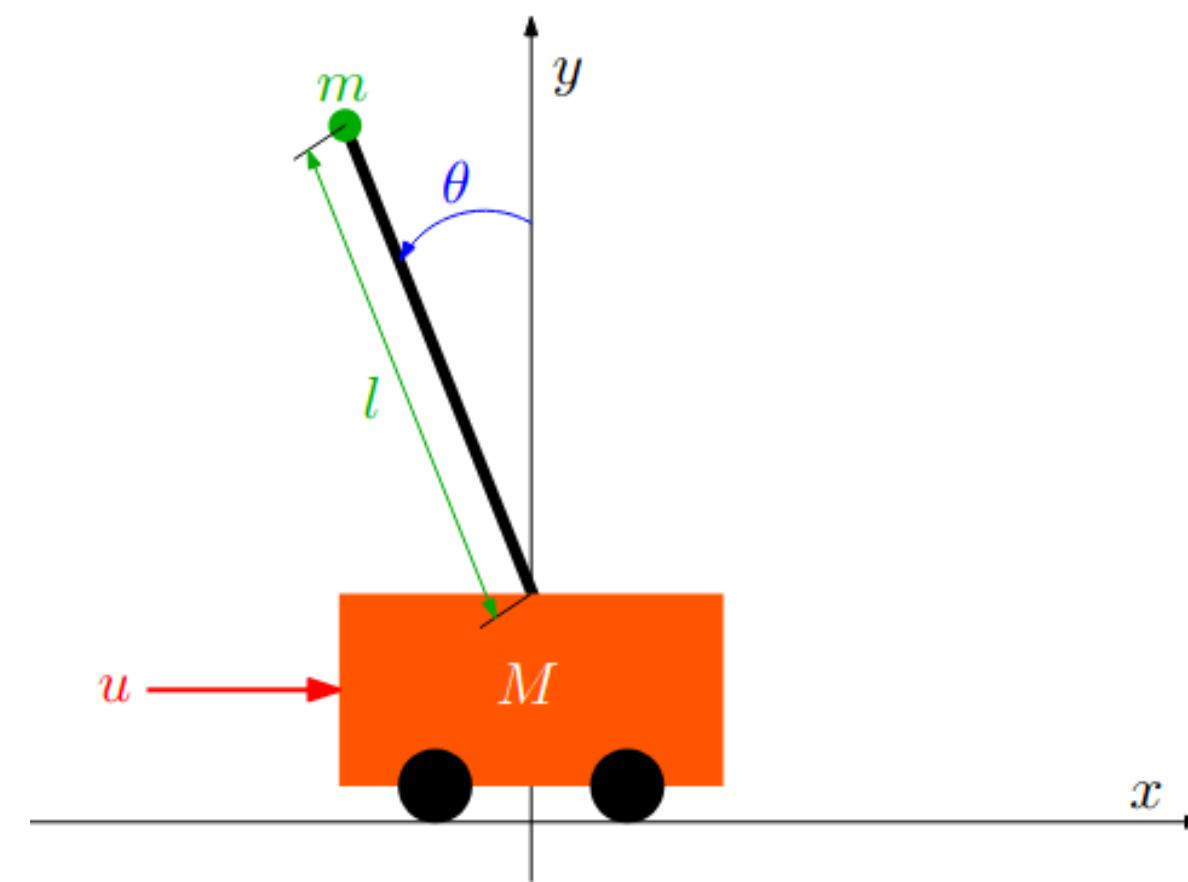


Time-triggered vs. Event-triggered schemes in a networked control system

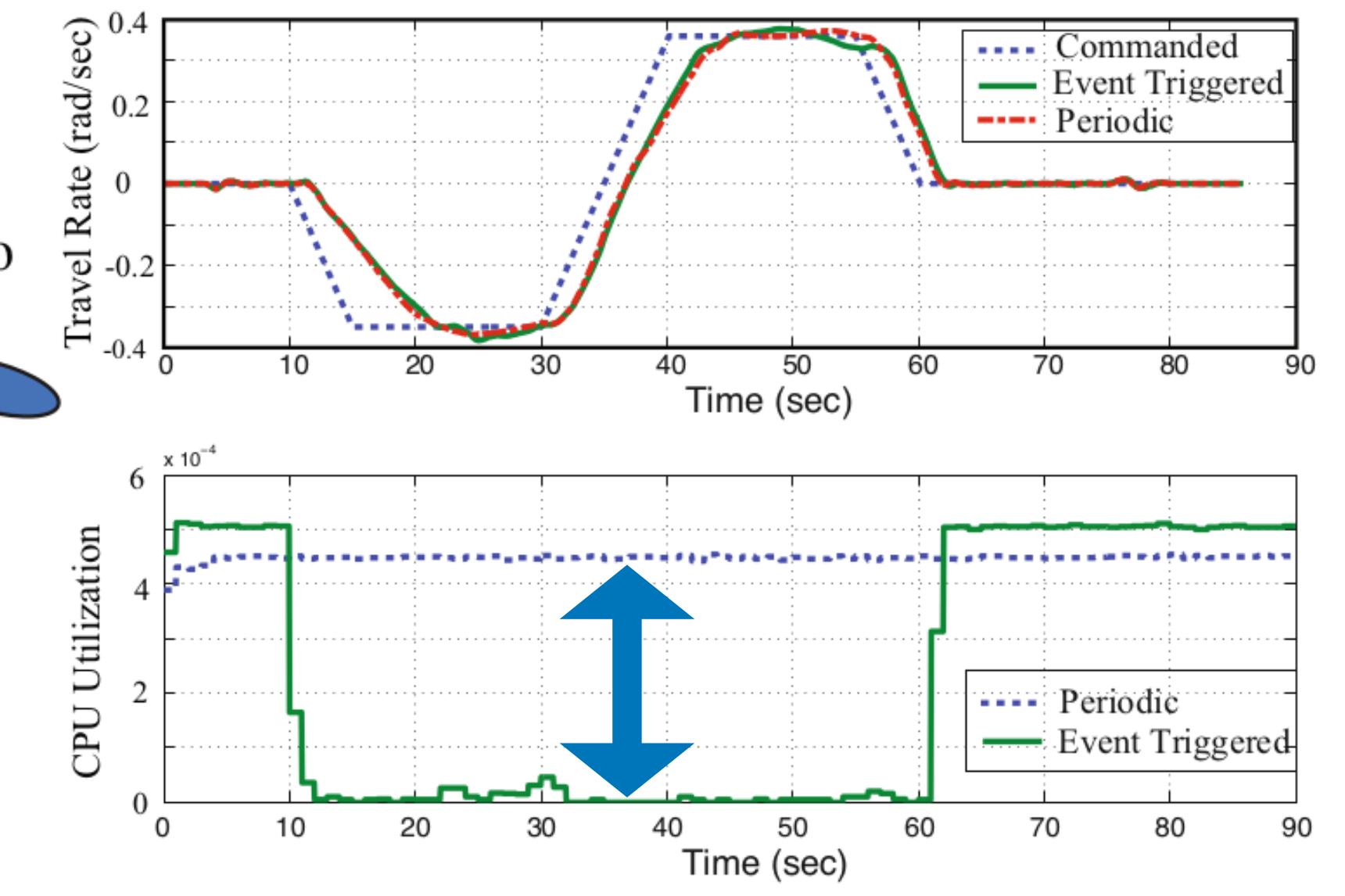
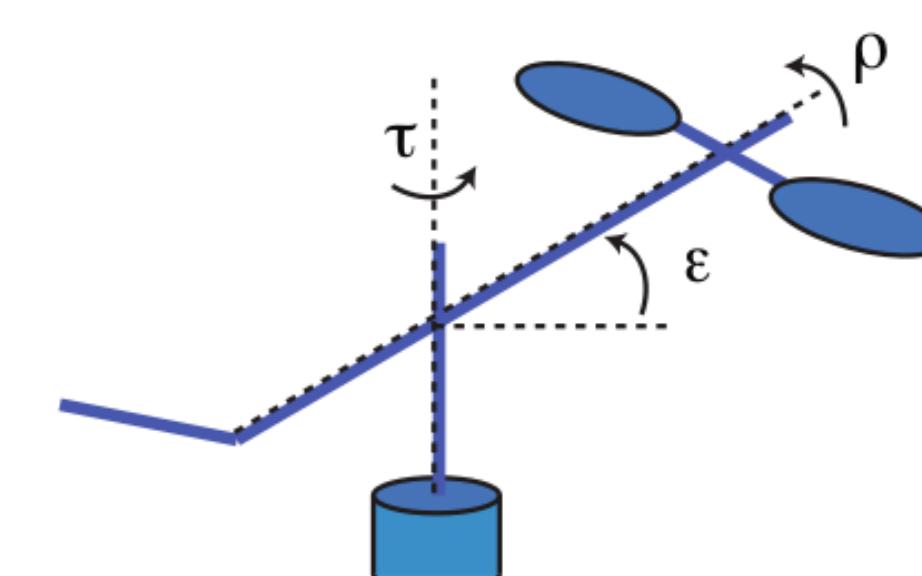
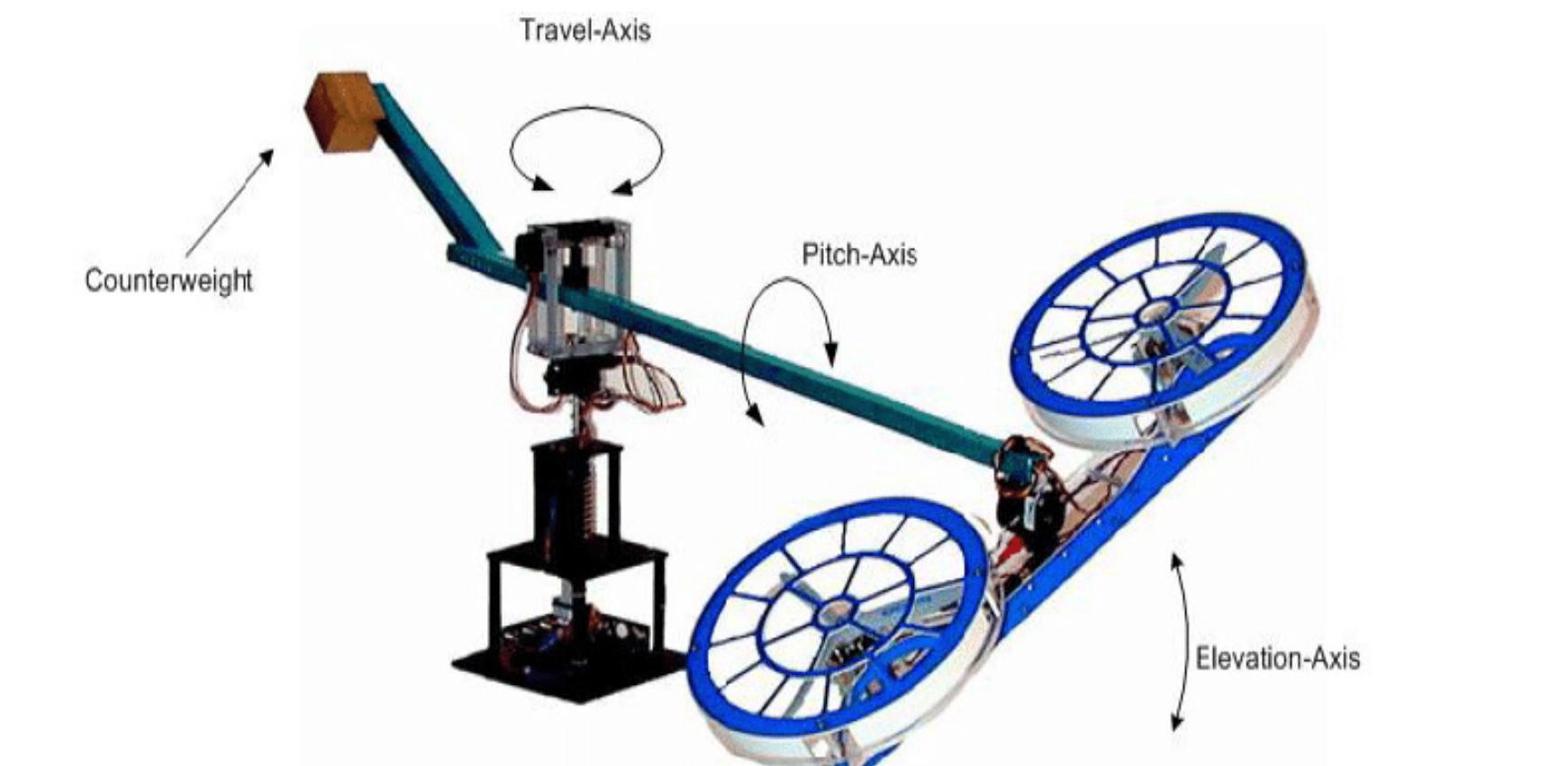
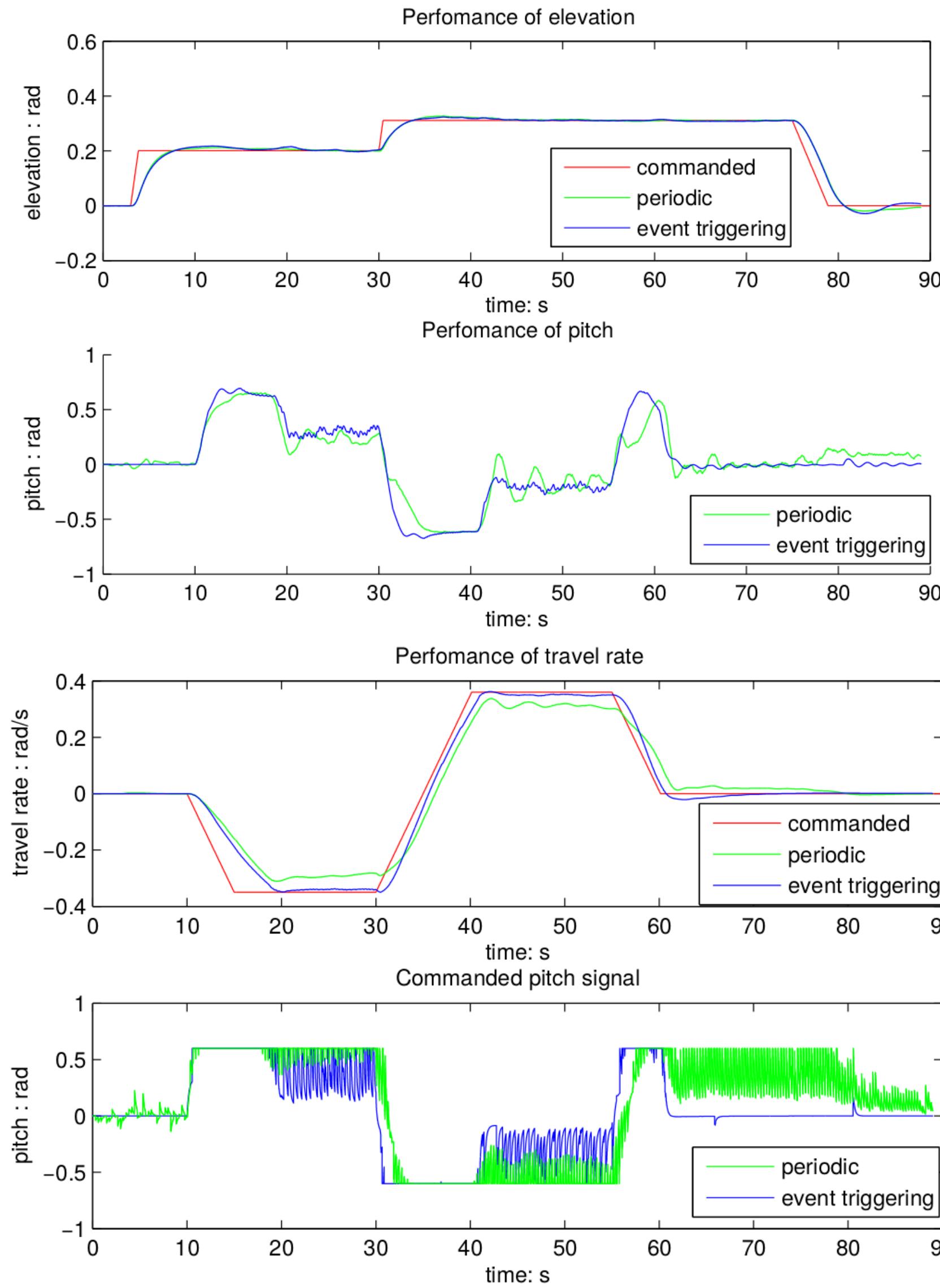
Event-Triggered vs. Time-Triggered Scheme: Inverted Pendulum

Compare the sampling instances:

- Time-triggered: 10000 samples
- Event-triggered: 3926 samples
- ~60% reduction in sampling for similar performance



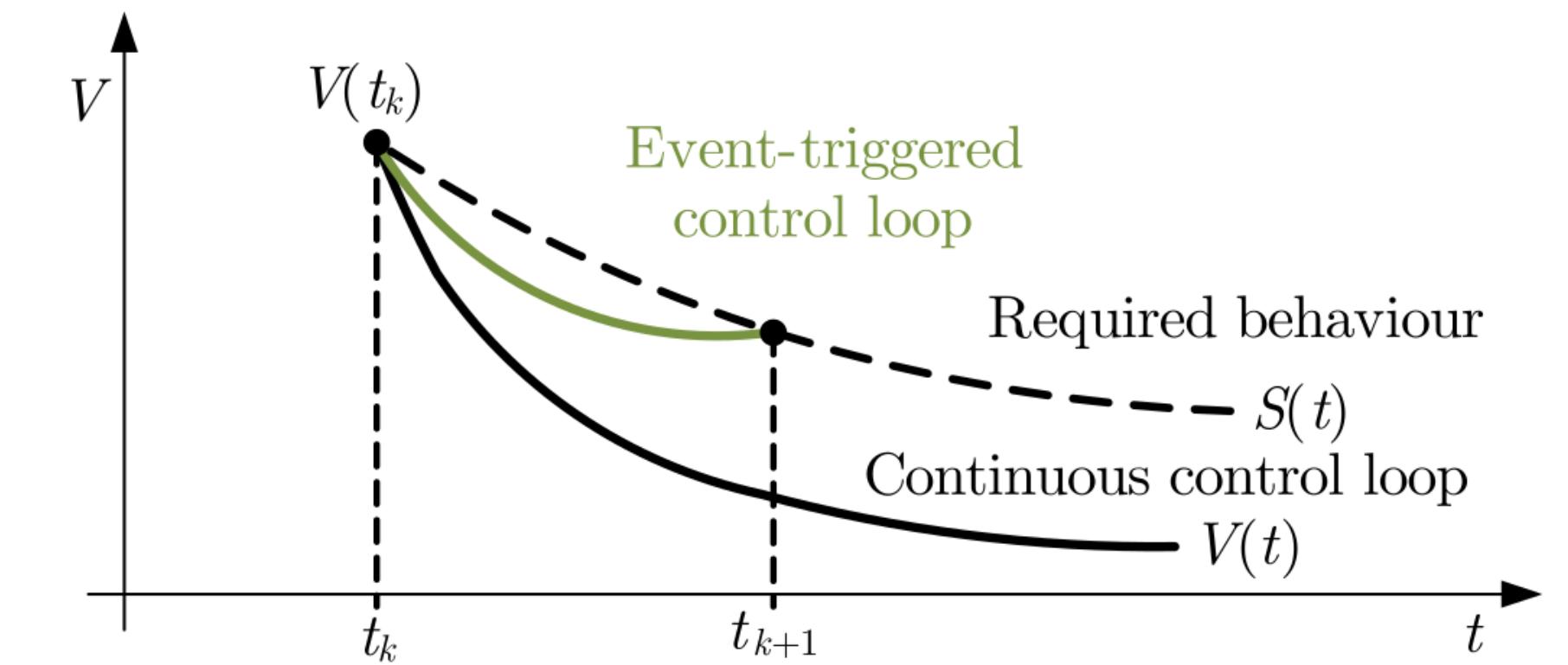
Event-Triggered vs. Time-Triggered Scheme: 3DOF Helicopter



Asynchronous Triggering Schemes

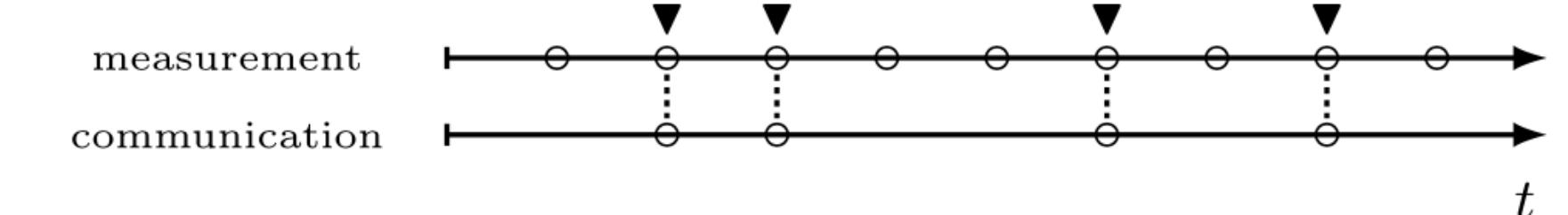
Event-triggered:

- Deterministic triggering functions: events are often deterministically defined with thresholds on system variables, e.g., the states of the system, estimation error, cost value, etc.
- Stochastic triggering functions: if events have stochastic nature, the triggering times are defined in probabilistic fashion
- Mixed triggering functions: events are triggered partly deterministically and partly stochastically



Periodic event-triggered:

- Continuous monitoring of system states to check when an event is triggered might not always be possible!
- Events are not continuously monitored but periodically, and if triggering condition is met, the action will be performed
- Often accompanied by maximum inter-update deadline, i.e., at least one update over a given interval even if event is not triggered



(a) Event-triggered control.

Event-triggered Control: Continuous-Time Linear Systems

Consider a continuous LTI system

$$\dot{x}(t) = Ax(t) + Bu(t), \quad u(t) = Lx(t), \quad x \in \mathbb{R}^n, \quad u \in \mathbb{R}^m$$

- $L \in \mathbb{R}^{m \times n}$ is a stabilising gain making the closed-loop system asymptotically stable!

Let $\{\tau_1, \tau_2, \tau_3, \dots\}$ be the sequence of executing the control inputs, i.e.,

$$u(t) = Lx(\tau_i), \quad \forall t \in [\tau_i, \tau_{i+1})$$

Define the error measure:

$$e(t) = x(\tau_i) - x(t), \quad \forall t \in [\tau_i, \tau_{i+1})$$

$\forall t \in [\tau_i, \tau_{i+1})$ the closed-loop system:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + BLx(\tau_i) \\ &= Ax(t) + BLx(t) - BLx(t) + BLx(\tau_i) \\ &= (A + BL)x(t) + BL(x(\tau_i) - x(t)) \\ &= (A + BL)x(t) + BLe(t)\end{aligned}$$

Lyapunov Stability for Linear Systems

Consider a continuous LTI system

$$\dot{x}(t) = Ax(t) + Bu(t), \quad u(t) = Lx(t), \quad x \in \mathbb{R}^n, \quad u \in \mathbb{R}^m$$

- $L \in \mathbb{R}^{m \times n}$ is a stabilising gain making the closed-loop matrix $A + BL$ stable!

Definition (Positive definite function): A continuous function $V: \mathbb{R}^n \rightarrow \mathbb{R}$ is called positive definite if it satisfies

1- $V(0) = 0$

2- $V(x) > 0, \forall x \neq 0, x \in \mathbb{R}^n$

Lyapunov Stability Theorem:

For the given linear closed-loop system $\dot{x}(t) = (A + BL)x(t)$, if exists a positive definite function $V: \mathbb{R}^n \rightarrow \mathbb{R}$ s.t.

$$\dot{V} = \frac{dV}{dx} \frac{dx}{dt} = \frac{dV}{dx}((A + BL)x(t)) = -W(x) \leq 0$$

1. if $W(x) \geq 0$, i.e., positive semi-definite, the equilibrium point is Lyapunov stable.
2. If $W(x) > 0$, i.e., positive definite, the equilibrium is asymptotically stable.

Event-triggered Control

Lemma: For a linear time-invariant system $\dot{x}(t) = Ax(t) + Bu(t)$, $u(t) = Lx(t)$ where L is a stabilising gain and $A + BK$ is a stable matrix, a Lyapunov function $V: \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$ exists, such that for some $\{\underline{a}, \bar{a}, a, g\} \in \mathbb{R}_{>0}$

$$\underline{a}|x|^2 \leq V(x) \leq \bar{a}|x|^2$$

$$\frac{dV}{dx}(Ax + BLx + BLe) \leq -a|x|^2 + g|e||x|$$

Define the constant positive parameter σ such that $-a + g\sigma < -b$, and $b > 0$. If it is guaranteed that:

$$|e| \leq \sigma|x|$$

We then have:

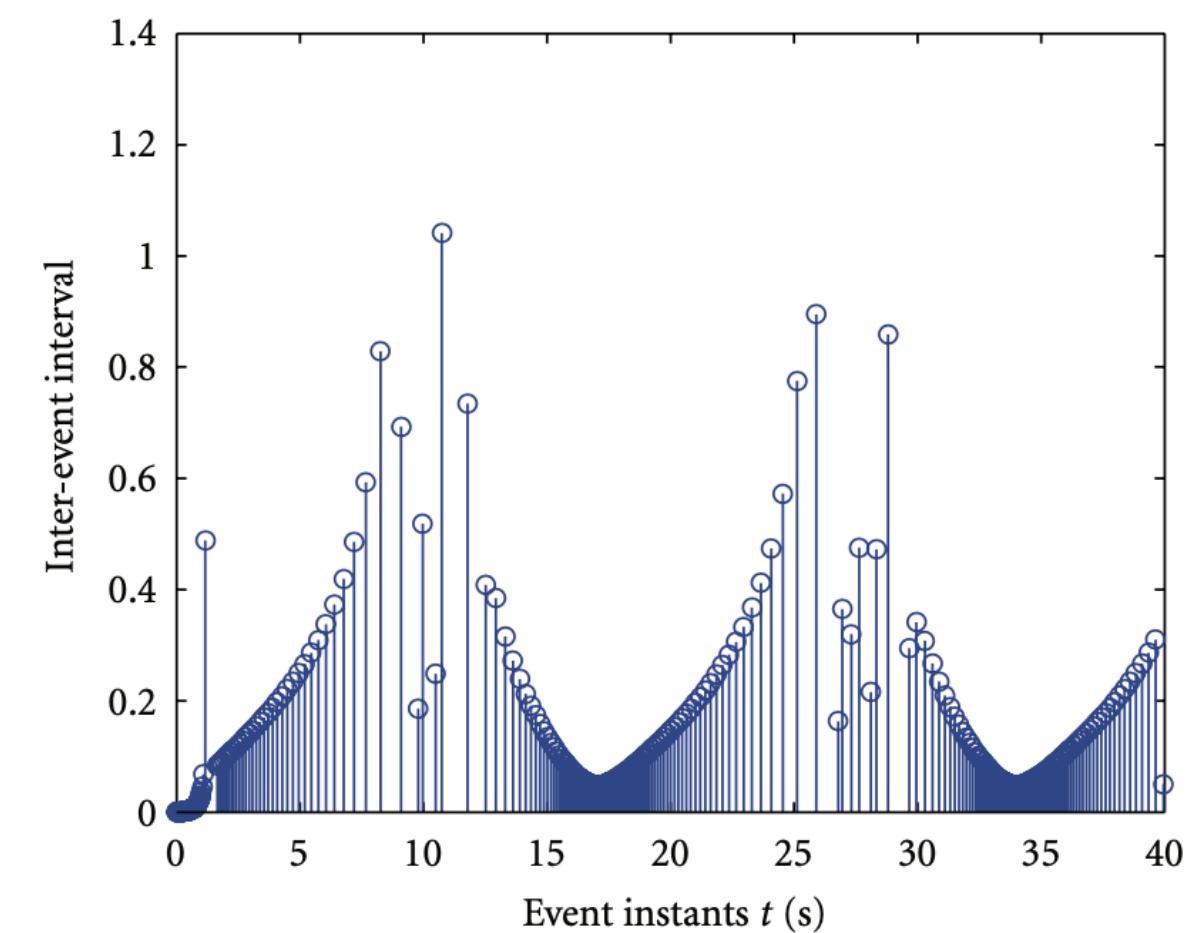
$$\frac{dV}{dt} \leq -a|x|^2 + g|e||x| \leq -a|x|^2 + g\sigma|x|^2 \leq (-a + g\sigma)|x|^2 < -b|x|^2$$

 We can execute the control input when the error value $|e|$ reaches to $\sigma|x|$ which guarantees asymptotic stability of the closed-loop control system: \rightarrow **we call $|e| = \sigma|x|$ a triggering law!**

Inter-event Times and Zeno Behavior

Inter-event time: the temporal duration $\tau_{i+1} - \tau_i$ between two consecutive triggering times is called inter-event time.

- For time-triggered/periodic schemes, the inter-event times are known and constant
- For event-triggered scheme, the inter-event times are dynamic, not known in advance, sometimes even random!



Zeno behavior: If an infinite number of events are triggered within a finite time

- Implementation of such control input policy on a digital controller is not possible due to limited sampling rate
- The triggering law should be designed carefully to avoid Zeno behavior!

Corollary: For a LTI control system $\dot{x}(t) = Ax(t) + Bu(t)$, let $u(t) = Lx(t)$ be a control law ensuring the closed loop system is asymptotically stable. Let the control signals be executed according to the event-triggered law $|e| \geq \sigma |x|$. Then the inter-event times $\tau_{i+1} - \tau_i$, $i \in \mathbb{N}$ resulting from the event-triggered execution are lower-bounded.

Proof of Corollary

Event-triggered Control: Discrete Time Linear Systems

Consider a discrete LTI system

$$x(t+1) = Ax(t) + Bu(t), \quad x \in \mathbb{R}^n, \quad u \in \mathbb{R}^m, \quad t \in \mathbb{N}_0$$

- Pair (A, B) is controllable, i.e., the matrix $(B, AB, A^2B, \dots, A^{n-1}B)$ has full rank.

Consider a triggering function $\xi : \mathbb{R}^n \rightarrow \mathbb{R}$, and define the event-triggered control input $u(t)$ as

$$u(t) = \begin{cases} Lx(t) & \text{if } \xi(x(t)) \geq 0 \\ u(t-1) & \text{if } \xi(x(t)) < 0 \end{cases}$$

L is the optimal feedback gain minimising the LQ cost function

$$J = \sum_{t=0}^T [x(t)^\top Q x(t) + u(t)^\top R u(t)]$$

- $L = -(R + B^\top P B)^{-1} B^\top P A$
- P solves the Algebraic Riccati equation: $P = Q + A^\top (P - PB(R + B^\top P B)^{-1} B^\top P)A$

Given $u(t)$, closed-loop system: $x(t+1) = \begin{cases} (A + BL)x(t) & \text{if } \xi(x(t)) \geq 0 \\ Ax(t) + BLx(\bar{t}) & \text{if } \xi(x(t)) < 0 \end{cases}$

- $\bar{t} \in \{0, 1, \dots, t-1\}$: time index of last triggered event!

Event-triggered Control

Define the Lyapunov function $V(x(t)) = x(t)^\top Px(t)$, $P > 0$, and compute $\Delta V = V(x(t+1)) - V(x(t))$:

- **Control input held constant:**

$$\Delta V_h = (Ax(t) + BLx(\bar{t}))^\top P (Ax(t) + BLx(\bar{t})) - V(x(t))$$

- **Control input updated:**

$$\Delta V_u = ((A + BL)x(t))^\top P ((A + BL)x(t)) - V(x(t)) < 0 \quad (\text{closed-loop system is stable})$$

Define the triggering function ξ as

$$\xi(x(t)) = \Delta V_h(x(t)) - \sigma \Delta V_u(x(t)), \quad \sigma \in (0,1]$$

- If $\xi(x(t)) \geq 0 \rightarrow \Delta V_u(x(t)) < 0$
- If $\xi(x(t)) < 0 \rightarrow \Delta V_h(x(t)) - \sigma \Delta V_u(x(t)) < 0 \rightarrow \Delta V_h(x(t)) < \sigma \Delta V_u(x(t)) < 0$

⇒ Triggering function ξ is stabilising as it ensures a monotonically decreasing Lyapunov function $V(x(t))$ along the trajectory of $x(t)$

⇒ σ is a tuning parameter and determines the triggering rate!

Event-triggered Control: A numerical Example

Consider the LTI control system

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -2x_1 + 3x_2 + u\end{aligned}$$

Controlled by the stabilising control inputs $u = x_1 - 4x_2$. Find a triggering value σ such that control input executions are governed by the law $|e| > \sigma|x|$ and the closed-system remains asymptotically stable.

Solution: Define Lyapunov function $V(x) = x^T Px > 0$ where $P = \begin{pmatrix} 1 & 0.25 \\ 0.25 & 1 \end{pmatrix}$ and $x = [x_1, x_2]^T$. We can derive

$$\frac{dV}{dt} = -x^T Q x + e^T L^T B^T P x + x^T P B L e, \quad Q = \begin{pmatrix} 0.5 & 0.25 \\ 0.25 & 1.5 \end{pmatrix}$$

Event-triggered closed-loop system asymptotically stable: $\frac{dV}{dt} \leq -a|x|^2 + g|e||x|$

- Selecting $a = \lambda_{min}(Q) = 1 - \frac{\sqrt{5}}{4}$ and $g = |L^T B^T P + P B L| = 8$ ensure the Lyapunov inequality holds!
- Since $\sigma g < a$ should hold, we should select $\sigma < 0.0551$. Selecting $\sigma = 0.05$ results in triggering instances whenever $|e| > 0.05|x|$.

References

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