

Trading loss against delay in Networked Control Systems

Rainer Blind, Frank Allgöwer

ist Institute for Systems Theory and Automatic Control University of Stuttgart, Germany

 $\mathbf{\Phi}$

DFG SPP 1305: Control Theory of Digitally Networked Dynamical Systems

Capacity Sharing Workshop Stuttgart, 13. October 2011

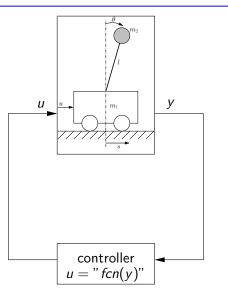


Feedback Control

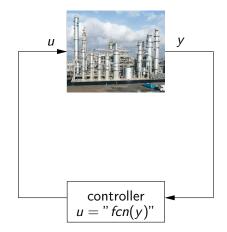


Show nice video http://www.youtube.com/watch?v=cyN-CRNrb3E http://www.youtube.com/watch?v=Ep21NMic_fk http://www.youtube.com/watch?v=B6vr1x6KDaY

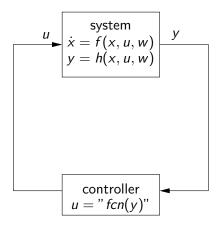
Control System



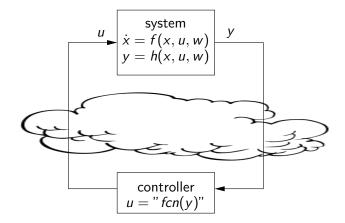
Control System



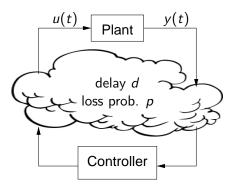
Control System







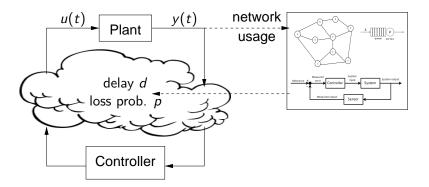
Loss and Delay in Networked Control Systems



Loss and Delay

- In control literature: assumed to be fixed.
- In communication literature: depends on network usage.

Loss and Delay in Networked Control Systems



Loss and Delay

- In control literature: assumed to be fixed.
- In communication literature: depends on network usage.

Outline





- 2 Problem Setup
- 3 Comparing Time-Triggered and Event-Based Control
- Trading Loss Against Delay
- 5 Outlook: Scalar System



Problem Setup



N Systems

$$\dot{x}(t) = u(t) + w(t),$$

where u(t) is the control input and w(t) noise.

Impulsive input

isto

$$u(t) = \sum_{k} -\delta(t-t_k)x(t),$$

each impulse resets the state x(t) to the origin.

Performance Measure (Cost)

$$J = \limsup_{M \to \infty} \frac{1}{M} \int_0^M \mathsf{E}[x(t)^2] dt$$



Problem Setup



Closed Loop System

$$t \neq t_k$$
: $u(t) = 0,$
 $\dot{x}(t) = w(t)$

$$t=t_k: \quad x(t_k)=0.$$



Performance Measure (Cost)

$$J = \limsup_{M \to \infty} \frac{1}{M} \int_0^M \mathsf{E}[x(t)^2] dt$$

Problem Setup



Closed Loop System

$$t \neq t_k$$
: $u(t) = 0,$
 $\dot{x}(t) = w(t)$

$$t=t_k: \quad x(t_k)=0.$$

When to reset? How to choose t_k ?

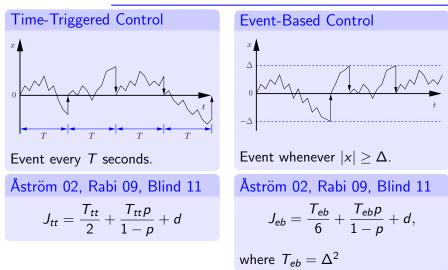
Performance Measure (Cost)

$$J = \limsup_{M \to \infty} \frac{1}{M} \int_0^M \mathsf{E}[x(t)^2] dt$$



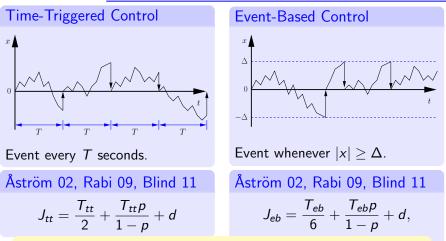
Comparing Time-Triggered and Event-Based Control





Comparing Time-Triggered and Event-Based Control

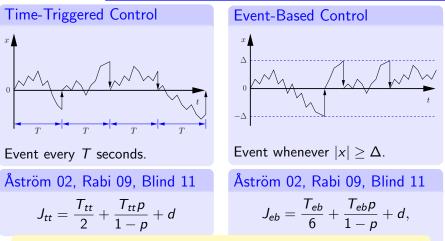




Literature: Event-based control is superior for NCS because fewer events are necessary to get the same performance.

Comparing Time-Triggered and Event-Based Control





Literature: Event-based control is superior for NCS because fewer events are necessary to get the same performance.

The traffic pattern is different.

Interevent Time Distribution of Event-Based Traffic



2

3

Interevent Time Distribution (Rabi 2009) The Probability Density Function (PDF) $f(t|\Delta)$ of the interarrival times T_{eb} of event-based control is given as: $f(t|\Delta) = \Delta \sqrt{\frac{2}{\pi t^3}} \sum_{k=-\infty}^{\infty} (4k+1)e^{-\frac{(4k+1)^2\Delta^2}{2t}}$.

Interevent Time Distribution of Event-Based Traffic



2

3

Interevent Time Distribution (Rabi 2009) The Probability Density Function (PDF) $f(t|\Delta)$ of the interarrival times T_{eb} of event-based control is given as: $f(t|\Delta) = \Delta \sqrt{\frac{2}{\pi t^3}} \sum_{k=1}^{\infty} (4k+1)e^{-\frac{(4k+1)^2\Delta^2}{2t}}$.

Theorem (Blind 2011)

The Palm-Khintchine Theorem holds: For $N \rightarrow \infty$, the arrival process of all agents together converges to a Poisson process.

Interevent Time Distribution of Event-Based Traffic



3

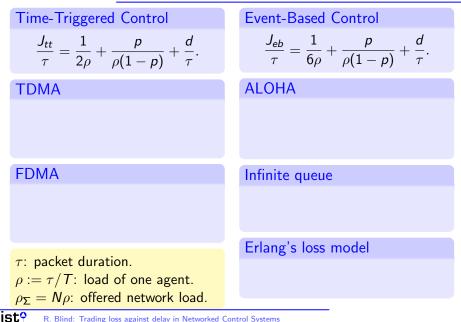
Interevent Time Distribution (Rabi 2009) f(t|1)The Probability Density Function (PDF) $f(t|\Delta)$ of the interarrival times 0.8 T_{eb} of event-based control is given as: 0.6 0.4 $f(t|\Delta) = \Delta \sqrt{\frac{2}{\pi t^3}} \sum_{k=-\infty}^{\infty} (4k+1)e^{-\frac{(4k+1)^2\Delta^2}{2t}}.$ 0.2 2

Theorem (Blind 2011)

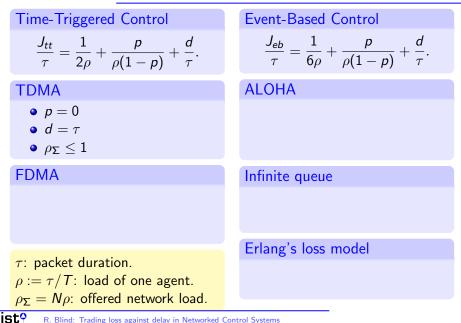
The Palm-Khintchine Theorem holds: For $N \to \infty$, the arrival process of all agents together converges to a Poisson process.

- 'Nice' traffic.
- Possible to use 'classic' theory to analyze NCS.

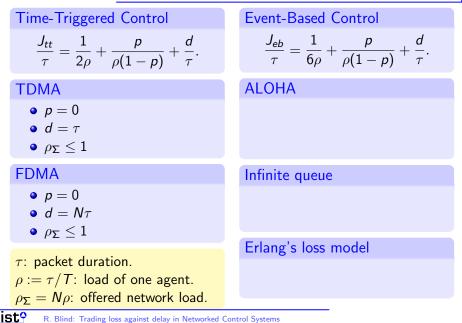
Trading Loss Against Delay



Trading Loss Against Delay



Trading Loss Against Delay



0

Trading Loss Against Delay

Time-Triggered Control

$$\frac{J_{tt}}{\tau} = \frac{1}{2\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

TDMA

- *p* = 0
- $d = \tau$
- *ρ*_Σ ≤ 1

FDMA

isto

- *p* = 0
- $d = N\tau$
- *ρ*_Σ ≤ 1

$$\begin{split} \tau\colon & \text{packet duration.} \\ \rho:=\tau/T\colon \text{load of one agent.} \\ \rho_{\Sigma}=N\rho: & \text{offered network load.} \end{split}$$

Event-Based Control

$$\frac{J_{eb}}{\tau} = \frac{1}{6\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

ALOHA

•
$$p_u = 1 - e^{-2\rho_{\Sigma}}$$
,
 $p_s = 1 - e^{-\rho_{\Sigma}}$
• $d_u = \tau$, $d_s = 1.5\tau$

Infinite queue

Erlang's loss model

0

Trading Loss Against Delay

Time-Triggered Control

$$\frac{J_{tt}}{\tau} = \frac{1}{2\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

TDMA

- *p* = 0
- $d = \tau$
- *ρ*_Σ ≤ 1

FDMA

isto

- *p* = 0
- $d = N\tau$
- *ρ*_Σ ≤ 1

$$\begin{split} \tau\colon & \text{packet duration.} \\ \rho:=\tau/T \colon \text{load of one agent.} \\ \rho_{\Sigma} = N\rho \colon & \text{offered network load.} \end{split}$$

Event-Based Control

$$\frac{J_{eb}}{\tau} = \frac{1}{6\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

ALOHA

•
$$p_u = 1 - e^{-2\rho_{\Sigma}}$$
,
 $p_s = 1 - e^{-\rho_{\Sigma}}$
• $d_u = \tau$, $d_s = 1.5\tau$

Infinite queue

•
$$p = 0$$

• $d = \frac{2-\rho_{\Sigma}}{2(1-\rho_{\Sigma})}\tau$

Erlang's loss model

0

Trading Loss Against Delay

Time-Triggered Control

$$\frac{J_{tt}}{\tau} = \frac{1}{2\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

TDMA

- *p* = 0
- $d = \tau$
- *ρ*_Σ ≤ 1

FDMA

isto

- *p* = 0
- $d = N\tau$
- *ρ*_Σ ≤ 1

$$\begin{split} \tau\colon & \text{packet duration.} \\ \rho:=\tau/T\colon \text{load of one agent.} \\ \rho_{\Sigma}=N\rho: & \text{offered network load.} \end{split}$$

Event-Based Control

$$\frac{J_{eb}}{\tau} = \frac{1}{6\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

ALOHA

•
$$p_u = 1 - e^{-2\rho_{\Sigma}}$$
,
 $p_s = 1 - e^{-\rho_{\Sigma}}$
• $d_u = \tau$, $d_c = 1.5\pi$

Infinite queue

•
$$p = 0$$

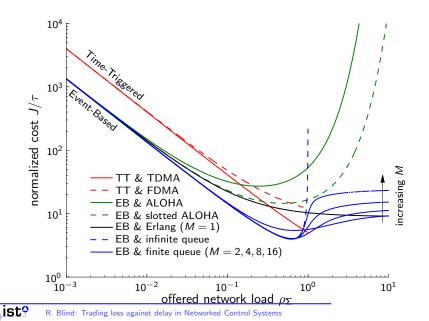
• $d = \frac{2-\rho_{\Sigma}}{2(1-\rho_{\Sigma})}\tau$

Erlang's loss model

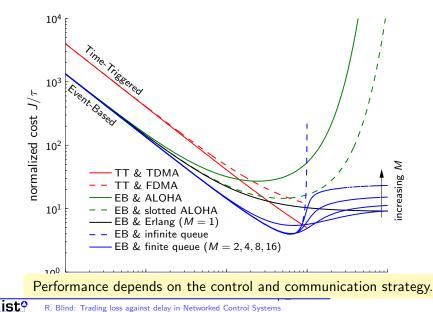
•
$$p = \frac{\rho_{\Sigma}}{1 + \rho_{\Sigma}}$$

• $d = \tau$

Normalized Cost for N = 8 agents



Normalized Cost for N = 8 agents



Outlook: Scalar System



Scalar System

$$\dot{x} = ax + u + w$$

Outlook: Scalar System

Scalar System

$$\dot{x} = ax + u + w$$

Theorem (Blind 2012)

Suppose, the scalar system is controlled by an event-based control scheme with bounds $\{\Delta_i\}$ and a packet loss probability *p*. Then the cost is

$$J = \frac{\sum_{i=0}^{\infty} p^{i} \int_{0}^{\Delta_{i}} \int_{0}^{t} x^{2} e^{a(x^{2}-t^{2})} dx dt}{\sum_{i=0}^{\infty} p^{i} \int_{0}^{\Delta_{i}} \int_{0}^{t} e^{a(x^{2}-t^{2})} dx dt}$$

Outlook: Scalar System

Scalar System

$$\dot{x} = ax + u + w$$

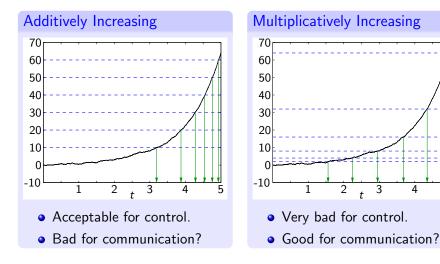
Theorem (Blind 2012)

Suppose, the scalar system is controlled by an event-based control scheme with bounds $\{\Delta_i\}$ and a packet loss probability *p*. Then the cost is

$$J = \frac{\sum_{i=0}^{\infty} p^{i} \int_{0}^{\Delta_{i}} \int_{0}^{t} x^{2} e^{a(x^{2}-t^{2})} dx dt}{\sum_{i=0}^{\infty} p^{i} \int_{0}^{\Delta_{i}} \int_{0}^{t} e^{a(x^{2}-t^{2})} dx dt}$$

How to choose the bounds $\{\Delta_i\}$?

How to Choose the Bounds?



isto

4

Summary and Outlook



Time-Triggered Control Deterministic traffic

$$\frac{J_{tt}}{\tau} = \frac{1}{2\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

Event-Based Control Poisson Traffic (for $N \rightarrow \infty$).

$$\frac{J_{eb}}{\tau} = \frac{1}{6\rho} + \frac{p}{\rho(1-\rho)} + \frac{d}{\tau}.$$

Conclusion

Performance depends on the control and communication scheme.

Summary and Outlook



Time-Triggered Control Deterministic traffic

$$\frac{J_{tt}}{\tau} = \frac{1}{2\rho} + \frac{p}{\rho(1-p)} + \frac{d}{\tau}.$$

Event-Based Control Poisson Traffic (for $N \rightarrow \infty$).

$$\frac{J_{eb}}{\tau} = \frac{1}{6\rho} + \frac{p}{\rho(1-\rho)} + \frac{d}{\tau}.$$

Conclusion

Performance depends on the control and communication scheme.

Outlook

- Scalar systems.
- Network instead of one bottleneck link.
- Send optimally.
- More realistic communication protocols.