

# 02\_NCS with packet losses and protocols

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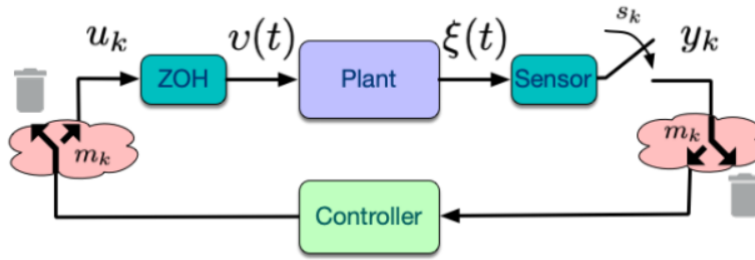
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## 1. Modeling of NCS with packet losses

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## 1.1. Assumptions

- No Delays
- Packet Losses Model:

- $$m_k = \begin{cases} 0, & \text{if no packet loss at time } k \\ 1, & \text{if packet is lost at time } k \end{cases}$$
- Assume two sides'  $m_k$  are synchronized

## 1.2. Controller Schemes

- Zero-order-hold:  $v(t) = u_k, t \in [s_k, s_{k+1})$

### 1.2.1. To-Hold Mechanism (event-driven)

$$u_k = \begin{cases} -\bar{K}x_k, & \text{if } m_k = 0 \\ u_{k-1}, & \text{if } m_k = 1 \end{cases}$$

### 1.2.2. To-Zero Mechanism (time-driven)

$$u_k = \begin{cases} -\bar{K}x_k, & \text{if } m_k = 0 \\ 0, & \text{if } m_k = 1 \end{cases}$$

## 1.3. Process Modeling

### Sampled-Data System, To-Zero Mechanism

With the controller  $u_k = -\bar{K}x_k$  and the "to zero" mechanism:

$$x_{k+1} = F_{m_k}^{cl}(h)x_k = \begin{cases} (F(h) - G(h)\bar{K})x_k & =: F_0^{cl}(h)x_k & \text{if } m_k = 0 \\ F(h)x_k & =: F_1^{cl}(h)x_k & \text{if } m_k = 1 \end{cases}$$

### Sampled-Data System, To-Hold mechanism

## 1.4. Models of Dropout Sequences

We need to model the "**discrete dynamics**" of the NCS:

$$m_{k+1} = f(m_k, m_{k-1}, \dots)$$

### Deterministic m Discrete Dynamics

Here we will use  $\omega$ -**automation**

An automaton is a 5-tuple  $M = (Q, \Sigma, \delta, Q_0, F)$ , where

- $Q$  is a finite or countable set of discrete states,
- $\Sigma$  is a finite or countable set of discrete inputs, the input alphabet,  $\Sigma$
- $\delta$  is the transition function,
- $Q_0 \subseteq Q$  is the set of **start states**, and
- $F \subseteq Q$  is the set of accept states.

### Stochastic m Discrete Dynamics

## 1.5. Model of Control Systems

### Deterministic Packet Loss

We consider a **Deterministic Special Case**: max.  $\delta$  consecutive losses

$$m \models (0^* \circ (10)^* \circ (110)^* \circ (1110)^* \circ \dots \circ \overbrace{(1 \dots 10)}^{\delta})^\omega.$$


So we can model a **closed-loop system-network** as:

- Discrete-time **switched system**

$$x_{I+1} = \tilde{F}^{cl}(h, h_l)x_I$$

- Packet-drops as extensions of sampling interval:

$$h_l \in \{h, 2h, \dots, (\delta + 1)h\}$$

- State  $x_l := x(\bar{s}_l)$ ,  $\bar{s}_{l+1} = \bar{s}_l + h_l$
-  *So we transfer deterministic packet loss to time-varying system*
  - With the "to zero" strategy:

$$x_{l+1} = \left( e^{Ah_l} - e^{A(h_l-h)} \int_0^h e^{As} B\bar{K} ds \right) x_l$$

Here the actual process is:

- First calculate after interval  $h$ :  $x'_l = \left( e^{Ah} - \int_0^h e^{As} B\bar{K} ds \right) x_l$
- For time interval:  $[t + h, t + h_l)$ , control signal become zeros, so we have  $x_{l+1} = e^{A(h_l-h)} x'_l$

### Stochastic Packet Loss

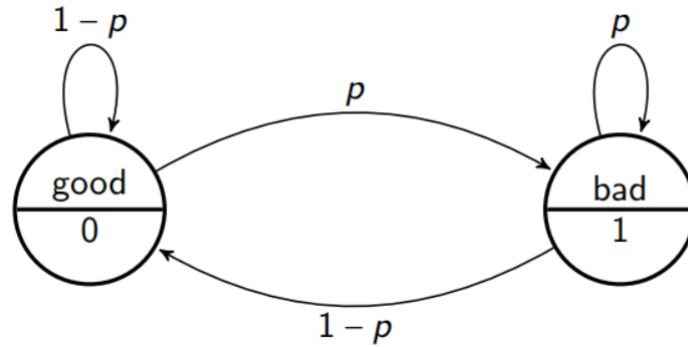
Here we use **Markov Decision Process** to model it

A Markov Decision Process (MDP) is a tuple  $(\Sigma, \Gamma, Q, q_0, \delta, H)$ , where:

- $\Sigma$  is an input alphabet (finite non-empty set of symbols);  $\Gamma$  is an output alphabet (possibly, and often a subset of  $\mathbb{R}$ )
- $Q$  is a finite, non-empty, set of states;
- $q_0 \in \mathbb{P}(Q)$  is the initial probability distribution of the states;
- $\delta : Q \times \Sigma \times Q \rightarrow [0, 1]$  is the state-transition relation;

- $H : Q \times \Sigma \times Q \rightarrow \Gamma$  is the output map or reward function.

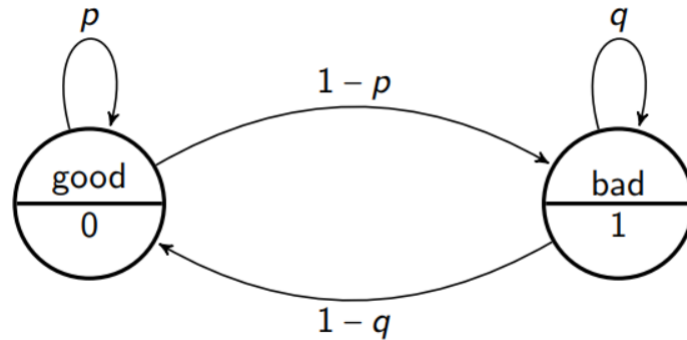
### Bernoulli Model



Finite State  $m_k \in \{0, 1\}$

$$\mathbb{P}(m_k = 1) = p$$

### Gilbert-Elliot Model



Finite State  $m_k \in \{0, 1\}$

$$\mathbb{P}(m_k = 0 \mid m_{k-1} = 0) = p_{00} = p, \mathbb{P}(m_k = 1 \mid m_{k-1} = 0) = p_{01} = 1 - p$$

$$\mathbb{P}(m_k = 1 \mid m_{k-1} = 1) = p_{11} = q, \mathbb{P}(m_k = 0 \mid m_{k-1} = 1) = p_{10} = 1 - q$$

## 2. Stability of NCS with Packet Loss

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## 2.1. Deterministic Packet Loss

### Total Model

$$h_l \in \{h, 2h, \dots, (\delta + 1)h\}$$

$$x_l := x(\bar{s}_l), \bar{s}_{l+1} = \bar{s}_l + h_l$$

- **we transfer deterministic packet loss to time-varying system**
  - With the "to zero" strategy:

$$x_{l+1} = \left( e^{Ah_l} - e^{A(h_1-h)} \int_0^h e^{As} B \bar{K} ds \right) x_l$$

### Stability Analysis

**Bounded Inter-sample Gain + Discrete-time system stability  $\Rightarrow$  Sampled-data stability**

Because:

$$h_l \in \mathcal{H} \Rightarrow h_l < \infty \Rightarrow \text{Bounded Inter-sample Gain}$$

## 2.2. Stochastic Stability Notions

Mean Square Stability(MSS), Exponentially MSS (EMSS), Uniformly EMSS(UEMSS)

A stochastic system is said to be **mean square stable (MSS)** if for all  $x_0$  and all  $m_{-1}$  :

$$\lim_{k \rightarrow \infty} \mathbb{E} \left[ \|x_k\|^2 \right] = 0$$

If furthermore:

- $\mathbb{E} \left[ \|x_k\|^2 \right] \leq c \rho^k \|x_0\|^2$  for some  $c \geq 0$  and  $\rho \in [0, 1)$  it is said to be **Exponentially MSS (EMSS)**,
- if  $c, \rho$  are independent of  $x_0$  and  $m_{-1}$ , it is said to be **Uniformly EMSS (UEMSS)**

*MSS: The final state's norm has a average 0*

*EMSS: The state's norm 's average is bounded by exponential functions*

*UEMSS: The state's norm's average is bounded by an unique exponential function*

## Stochastic Stability (SS)

A stochastic system is said to be **stochastically stable (SS)** if for all  $x_0$  and all  $m_{-1}$  :

$$\mathbb{E} \left[ \sum_{k=0}^{\infty} \|x_k\|^2 \right] < \infty$$

*The average energy of the systems is finite*

## Almost Surely Stable (ASS)

A stochastic system is said to be **almost surely stable (ASS)** if for all  $x_0$  and all  $m_{-1}$  :

$$\begin{aligned} \mathbb{P} [\lim_{k \rightarrow \infty} \|x_k\|] &= 0. \\ &= P [\lim_{k \rightarrow \infty} \|x_k\| = 0] = 1 \end{aligned}$$

## Relation

For Markovian Jump Linear Systems (MJLS) the following holds:

$$MSS \Leftrightarrow SS \Leftrightarrow EMSS \Leftrightarrow UEMSS \Rightarrow ASS$$

*ASS means the system has countable number of situation that may lead to a non-stable sequence (not converge to 0)*

*MSS means the system has countable number of situation that may lead to a non-stable sequence and the non-stable sequence has a bounded deviation from fixed point.*

## 2.3. Stability of NCS with Stochastic Packet Loss

### Overall Model

$$x_{k+1} = F_{m_k}^{cl}(h)x_k, \quad \mathbb{P}(m_k = j \mid m_{k-1} = i) = p_{ij}$$

**Switched system**,  $m$  controls **switching stochastically** (Markovian) : **Markovian Jump Linear System**

### Stability Analysis (MJLS MSS)

A Markovian Jump Linear System:

$$x_{k+1} = A_{m_k}x_k, \quad \mathbb{P}(m_k = j \mid m_{k-1} = i) = p_{ij}$$

is **Mean Square Stable (MSS)** if there **exist**  $P_i > 0, i = 0, \dots, N$  that satisfy **any of** the following conditions:

- $P_i - A_i^T \left( \sum_{j=0}^N p_{ij} P_j \right) A_i > 0$  for all  $i = 0, \dots, N$
- $P_j - A_j \left( \sum_{i=0}^N p_{ij} P_i \right) A_j^T > 0$  for all  $j = 0, \dots, N$
- $P_i - \sum_{j=0}^N p_{ij} A_j^T P_j A_j > 0$  for all  $i = 0, \dots, N$
- $P_j - \sum_{i=0}^N p_{ij} A_i P_i A_i^T > 0$  for all  $j = 0, \dots, N$

**Note:** In the case we consider  $N = 1$ , as  $m_k \in \{0, 1\}$ .

**Understanding:**

$$P_i - \sum_{j=0}^N p_{ij} A_j^T P_j A_j > 0 \quad \text{for all } i = 0, \dots, N$$

$\Rightarrow$  The Lyapunov function  $V(x_k, m_{k-1}) = x_k^T P_{m_{k-1}} x_k$  satisfies:

$$\mathbb{E}[V(x_{k+1}, x_k) \mid x_k] < V(x_k, m_{k-1}), \forall x_k \neq 0$$



## For Bernoulli Probabilistic Transitions

$$x_{k+1} = A_{m_k}(h)x_k, \quad \mathbb{P}(m_k = j \mid m_{k-1} = i) = p_j$$

there exists  $P > 0$  such that:

$$\begin{aligned} P_0 - (1-p)A_0^T P_0 A_0 - pA_1^T P_1 A_1 &> 0 \\ P_1 - pA_1^T P_1 A_1 - (1-p)A_0^T P_0 A_0 &> 0 \end{aligned}$$

$\Leftrightarrow$  there exists  $P > 0$  such that:

$$P - (1-p)A_0^T P A_0 - pA_1^T P A_1 > 0$$

## For Gilbert-Elliot Model

$$P_i - \sum_{j=0}^N p_{ij} A_j^T P_j A_j > 0, \forall i, j \in \{0, 1\}$$

$\Leftrightarrow$

$$\begin{aligned} P_0 - pA_0^T P_0 A_0 - (1-p)A_1^T P_1 A_1 &> 0 \\ P_1 - qA_1^T P_1 A_1 - (1-q)A_0^T P_0 A_0 &> 0 \end{aligned}$$

## For B-Model and GE-Model with To-Zero Mechanism

**MSS if and only if**

- (Bernoulli) there exists  $P > 0$  such that:

$$P - (1-p)F_0^{c/T} P F_0^{cl} - pF_1^{c/T} P F_1^{cl} > 0$$

- (Gilbert-Elliot) there exist  $P_0, P_1 > 0$  such that:

$$\begin{aligned} P_0 - pF_0^{c/T} P_0 F_0^{cl} - (1-p)F_1^{c/T} P_1 F_1^{cl} &> 0 \\ P_1 - qF_1^{c/T} P_1 F_1^{cl} - (1-q)F_0^{c/T} P_0 F_0^{cl} &> 0 \end{aligned}$$

where:

$$\begin{aligned} F(h) &:= e^{Ah}, \quad G(h) := \int_0^h e^{As} B ds \\ F_1^{cl} &:= F(h), \quad F_0^{cl} := (F(h) - G(h)\bar{K}) \end{aligned}$$

# Almost Sure Stability Theorem for Bernoulli Model

The system:

$$x_{k+1} = e^{Ah}x_k + \int_0^h e^{As}Bdsu_k$$

with the controller  $u_k = -\bar{K}x_k$ , "to zero" mechanism, and Bernoulli packet losses is ASS ( $\mathbb{P}[\lim_{k \rightarrow \infty} \|x_k\| = 0] = 1$ ) if there exists a  $V : \mathbb{R}^n \rightarrow \mathbb{R}^+$  such that  $c_1\|x\|^r \leq V(x) \leq c_2\|x\|^r$ , with  $r \geq 1, 0 < c_1 \leq c_2$ , satisfying:

- $V(F_0^{cl}x) \leq \lambda V(x)$  for  $\lambda \in [0, 1)$ ,
- $V(F_1^{cl}x) \leq LV(x)$  for some  $L \geq 0$ ,
- $\lambda^{(1-p)}L^p < 1$ .

*Can be partly increase when loss happen, however, the average effect of decrease and increase should smaller than 1*

## 3. NCS with communication constraints: Protocol

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