

On the input energy for state reachability of linear systems with packet losses

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- Introduction
- Problem statement
- The Controllability Gramian and reachability metrics
- Computation and main results
- Examples
- Conclusion and future work

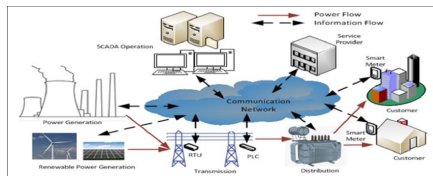
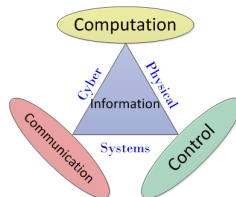


Figure : Cyber Physical Systems



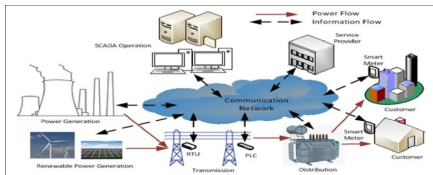
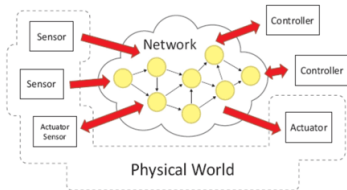
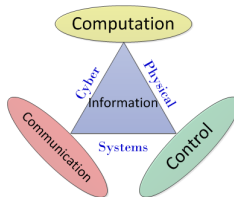


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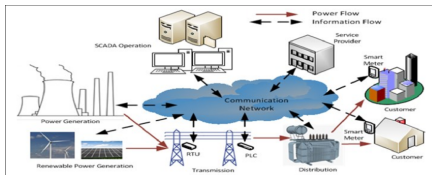
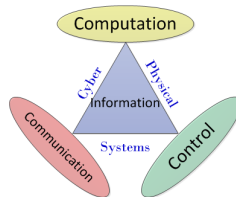
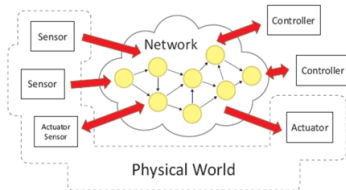


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- Packet loss in wireless communication: a common non-ideality.



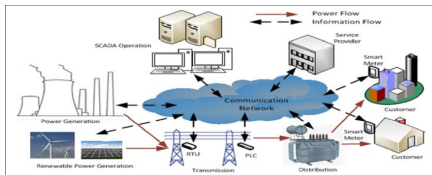
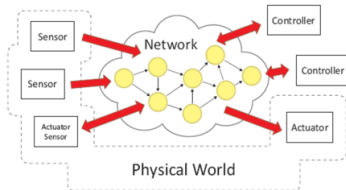
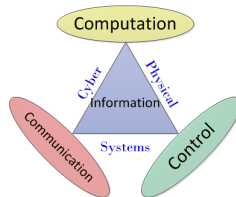


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- Packet loss in wireless communication: a common non-ideality.
- Greatly influences **controllability, observability, required control energy** etc.
- [Jungers, Kundu and Heemels]: **controllability and observability are decidable.**

- Discrete linear systems subject to data losses :
 $x(t+1) = Ax(t) + B\sigma(t)u(t)$ where

hybrid systems [Jungers, Kundu and Heemels]

$$x(t+1) = \begin{cases} Ax(t) + Bu(t), & \text{if } \sigma(t) = 1 \\ Ax(t), & \text{if } \sigma(t) = 0. \end{cases} \quad (1)$$

- σ is a signal which models the **packet loss**.

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- σ is a signal which models the **packet loss**.
- We say that a signal σ is **admissible** if it is allowed by some **automaton**.

Definition ([Jungers, Kundu and Heemels])

We say that the hybrid system is controllable, if for all admissible signals $\sigma : \mathbb{N} \rightarrow \{0, 1\}$, any initial state $x_0 \in \mathbb{R}^n$ and any final state $x_f \in \mathbb{R}^n$, there is an input signal u such that

$$x_{x_0, \sigma, u}(T) = x_f \text{ for some } T \in \mathbb{N}.$$

If $x_0 = 0$, we say that the system is reachable.

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We want to study how the **energy** for reachability is affected when there are **packet dropouts**.

- The Controllability Gramian for discrete linear system

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- The **least input energy** required to drive the state from $x(0) = x_0$ to $x(t+1) = x_f$

$$E(x_0, x_f, t) = (x_f - A^t x_0)^T W_t(A, B)^{-1} (x_f - A^t x_0). \quad (4)$$

- 1 The minimum eigenvalue of the controllability Gramian: $\lambda_{min}(W_t(A, B))$.
 - 2 The trace of the inverse of the controllability Gramian: $\text{tr}(W_t(A, B))^{-1}$.
 - 3 The determinant of the controllability Gramian: $\det(W_t(A, B))$.
- $\lambda_{min}(W_t(A, B)) \leftrightarrow$ maximum energy for reachability on the unit sphere.
 - $\text{tr}(W_t(A, B))^{-1} \leftrightarrow$ average energy for reachability on the unit sphere.
 - $\det(W_t(A, B)) \leftrightarrow$ volume of the ellipsoid that can be reached with the unit energy input from the origin.

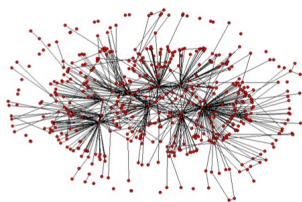


Figure : Complex networks

- [Summers et al.], [Pasqualetti et al.] Problem: optimal actuator placements in complex networks. To choose k number of actuators from a given set to maximize a controllability metric.

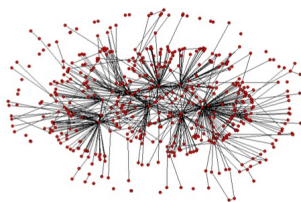


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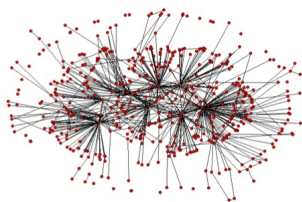


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- [Summers et al.]: Greedy algorithm and the sub-modularity properties of controllability metrics except the minimum eigenvalue of the Controllability Gramian were used to solve this combinatorial problem.
- [Pasqualetti et al.] obtained upper bounds on the minimum eigenvalue in terms of the number of actuators and the number of stable eigenvalues.

- Trade-offs between the input energy and the number of actuators were obtained for large scale networks.
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- [Olshovsky] obtained upper bounds on the minimum eigenvalue of the Controllability Gramian for linear time invariant systems using tools from potential theory.
- It was shown that if eigenvalues of A are clustered together, it requires more energy to control.

- Trade-offs between the input energy and the number of actuators were obtained for large scale networks.
- In specific, lower bounds on the number of actuators was obtained in terms of the fixed input energy [Pasqualetti et al.].
- [Olshesky] obtained upper bounds on the minimum eigenvalue of the Controllability Gramian for linear time invariant systems using tools from potential theory.
- It was shown that if eigenvalues of A are clustered together, it requires more energy to control.
- Our combinatorial problem is different than the one considered in these references and the techniques used therein are not applicable to our case.

- Given a **fixed amount of energy or average energy**, decide if the reachability problem is **feasible** for all admissible switching signals; and **identify switching signals** for which it is **infeasible**.

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- Provide **algorithms** to decide this question.
- **Quantify** the **performance** of **communication networks** using the energy required for control.

Definition

- An automaton is a directed labelled graph $G(V, E)$ with N_V nodes in V and N_E edges in E .
- An edge $(v, w, \sigma) \in E$ carries a label $\sigma \in \{0, 1\}$.
- A sequence $\sigma(0), \sigma(1), \dots$ is accepted by the graph G if there is a path in G carrying the sequence as the succession of labels on its edges.
- We denote by $\mathcal{L}(\mathcal{A})$ the set of all admissible switching sequences.

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Example ([Jungers, Kundo and Heemels])

- Consider a network where there can be at most 3 consecutive dropouts.

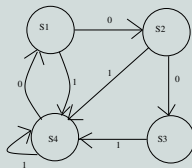


Figure : \mathcal{A}

- The controllability matrix associated with a signal σ at time t

$$C_{\sigma(t)}(A, B) = \begin{bmatrix} A^t B \sigma(0) & \cdots & A B \sigma(1) & B \sigma(t) \end{bmatrix}. \quad (5)$$

- The controllability Gramian for hybrid systems at time t with respect to the signal σ

$$W_{\sigma(t)} := \sum_{i=0}^t \sigma(t-i) A^i B B^T (A^T)^i \quad (6)$$

$$= C_{\sigma(t)}(A, B) C_{\sigma(t)}(A, B)^T. \quad (7)$$

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- $\bar{u} = C_{\sigma(t)}(A, B)^T (W_{\sigma(t)})^{-1} (x(t+1) - A^t x(0))$ is the minimum energy input which does the required state transfer.
- The minimum input energy required to drive the state from $x(0) = 0$ to $x(t+1) = x_f$ for a switching signal σ is

$$E_{\sigma}(\{0\}, \{x_f\}, t) = (x_f)^T W_{\sigma(t)}^{-1} (x_f). \quad (8)$$

- $\max_{x_0=0, x_f \in \mathbb{S}^n} \{E_\sigma(\{0\}, \{x_f\}, t)\} = \lambda_{\min}(W_\sigma(t))^{-1}$.
- $E(\{0\}, \mathbb{S}^n, t) := \max_{\sigma \in \mathcal{L}(\mathcal{A}), x_0=0, x_f \in \mathbb{S}^n} \{E_\sigma(\{0\}, \{x_f\}, t)\}$.

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- We are interested in computing the following quantities

$$\lambda(t) := \min_{\sigma} (\lambda_{\min}(W_{\sigma(t)})). \quad (9)$$

$$S_{\sigma}(t) := \operatorname{argmin}_{\sigma} (\lambda_{\min}(W_{\sigma(t)})). \quad (10)$$

$$\lambda := \lim_{t \rightarrow \infty} \lambda(t). \quad (11)$$

- $\lambda(t)^{-1} = E(\{0\}, \mathbb{S}^n, t)$.

Definition

Analogous to the previous definition, we define

$$\bar{S}_\sigma(t) := \operatorname{argmax}_\sigma(\operatorname{tr}(W_{\sigma(t)}^{-1})). \quad (12)$$

Let $\sigma^\dagger \in \bar{S}_\sigma(t)$.

$$\delta(t) := \max_\sigma(\operatorname{tr}(W_{\sigma(t)}^{-1})) = \operatorname{tr}(W_{\sigma^\dagger(t)}^{-1}). \quad (13)$$

When A is stable, we define

$$\delta := \lim_{t \rightarrow \infty} \delta(t). \quad (14)$$

Definition (partial order)

Given two switching signals σ_1 and σ_2 , we say that $\sigma_1 \preceq \sigma_2$ if $\sigma_1(i) = 1$ ($i \in \mathbb{Z}$), then $\sigma_2(i) = 1$ but not conversely. If the converse holds, then $\sigma_1 = \sigma_2$.

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Definition (Minimal signals)

We say that a signal σ is minimal, if there does not exist any other signal $\bar{\sigma}$ ($\bar{\sigma} \neq \sigma$) allowed by the automaton such that $\bar{\sigma} \preceq \sigma$. We denote by $M_{\sigma(t)}$ the set of all minimal signals defined from 0 to t .

Lemma

Suppose time t is given and $\sigma_1 \preceq \sigma_2$. Then

- 1 $\lambda_{\min}(W_{\sigma_1(t)}) \leq \lambda_{\min}(W_{\sigma_2(t)})$.
- 2 $\text{tr}(W_{\sigma_1(t)}^{-1}) \geq \text{tr}(W_{\sigma_2(t)}^{-1})$.

Reducing computations

Instead of all signals allowed by the automaton, we need to check the minimal signals to compute $\lambda(t)$ and $S_\sigma(t)$.

How to generate the minimal signals \rightarrow T -product lift of graphs

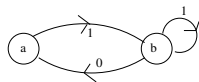


Figure : Automaton

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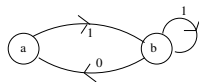


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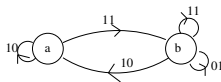


Figure : 2-lift

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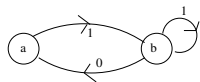


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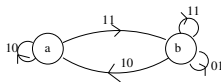


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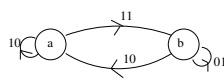


Figure : 2-lift reduction

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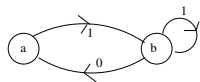


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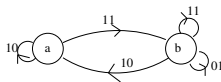


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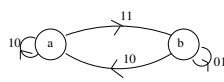


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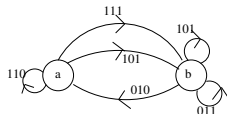


Figure : 3-lift

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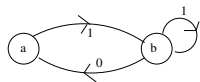


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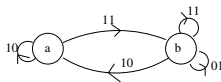


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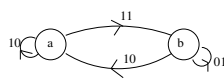


Figure : 2-lift reduction

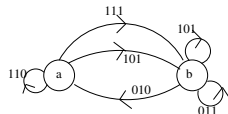


Figure : 3-lift

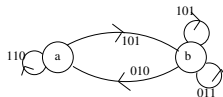


Figure : 3-lift reduction

How to generate the minimal signals \rightarrow T -product lift of graphs

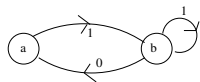


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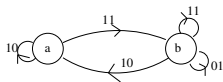


Figure : 2-lift

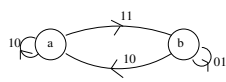


Figure : 2-lift reduction

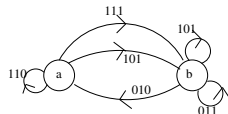


Figure : 3-lift

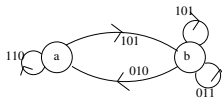


Figure : 3-lift reduction

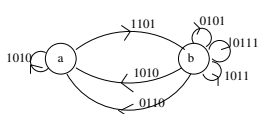


Figure : 4-lift

How to generate the minimal signals \rightarrow T -product lift of graphs

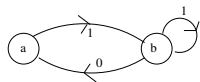


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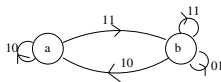


Figure : 2-lift

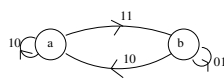


Figure : 2-lift reduction

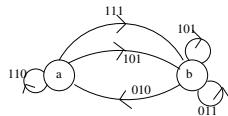


Figure : 3-lift

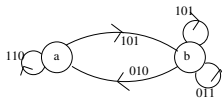


Figure : 3-lift reduction

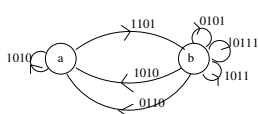


Figure : 4-lift

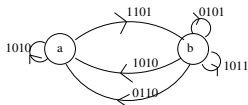


Figure : 4-lift reduction

- We assume that A is invertible and the system is controllable.

Definition

Let \mathcal{A} be an automaton with m nodes. Let σ be an admissible signal of length m . A signal σ is said to be m -minimal if it is a minimal signal with respect to all signals of length m allowed by \mathcal{A} .

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Let \mathcal{A} be an automaton with m nodes. Let σ be an admissible signal of length m . A signal σ is said to be m -minimal if it is a minimal signal with respect to all signals of length m allowed by \mathcal{A} .

Definition (sparse minimal signals)

Let σ be a minimal admissible signal allowed by an automaton \mathcal{A} having m nodes. Partition the time interval from 0 to t into blocks of length m . Then the signal σ is said to be sparse minimal signal if each block in its partition is m -minimal. We denote by $Sp_{\sigma(t)}$ the set of all sparse minimal signals from 0 to t .

Definition

We define the approximate of $\lambda(t)$ as follows

$$\lambda_{app}(t) := \min_{\sigma \in Sp_{\sigma(t)}} \{\lambda_{min}(W_{\sigma(t)})\} \quad (15)$$

Example

- Suppose at most 3 successive dropouts are allowed and $t = 7$. Thus $m = 4$.
- m -minimal signals are (1){0, 0, 0, 1}, (2){0, 0, 1, 0}, (3){0, 1, 0, 0}, (4){1, 0, 0, 0}.
- The sparse minimal signals are as follows:

{0, 0, 0, 1, 0, 0, 0, 1},	
{0, 0, 0, 1, 0, 0, 1, 0},	{0, 1, 0, 0, 0, 1, 0, 0},
{0, 0, 0, 1, 0, 1, 0, 0},	{0, 1, 0, 0, 1, 0, 0, 0}
{0, 0, 0, 1, 1, 0, 0, 0},	
{0, 0, 1, 0, 0, 0, 1, 0},	
{0, 0, 1, 0, 0, 1, 0, 0},	{1, 0, 0, 0, 1, 0, 0, 0}.
{0, 0, 1, 0, 1, 0, 0, 0}	

Examples: the number of signals vs time

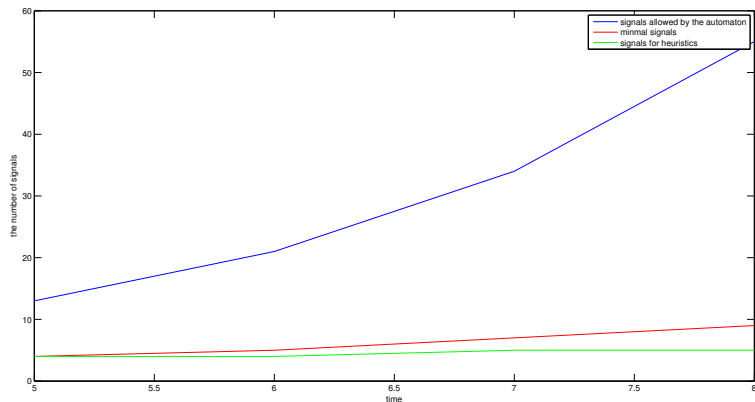


Figure : Automaton: no more than one consecutive dropouts

Examples: the number of signals vs time

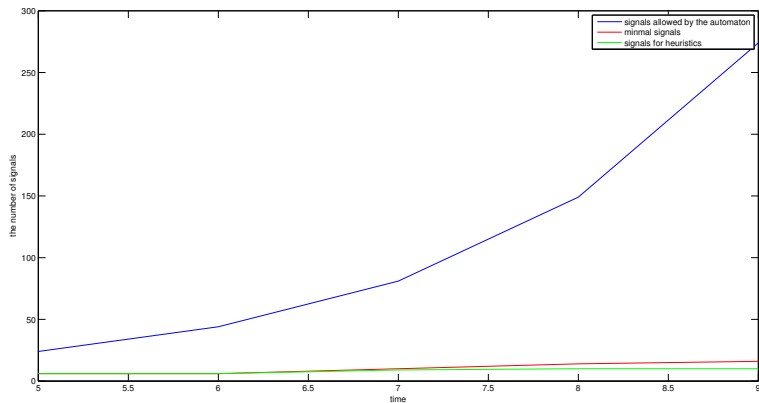


Figure : Automaton: no more than two consecutive dropouts

Examples: the number of signals vs time (logarithmic plot)

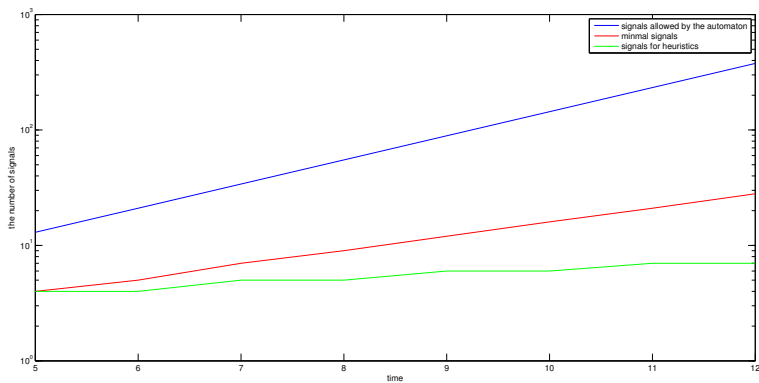


Figure : Automaton: no more than one consecutive dropouts

Examples: the number of signals vs time (logarithmic plot)

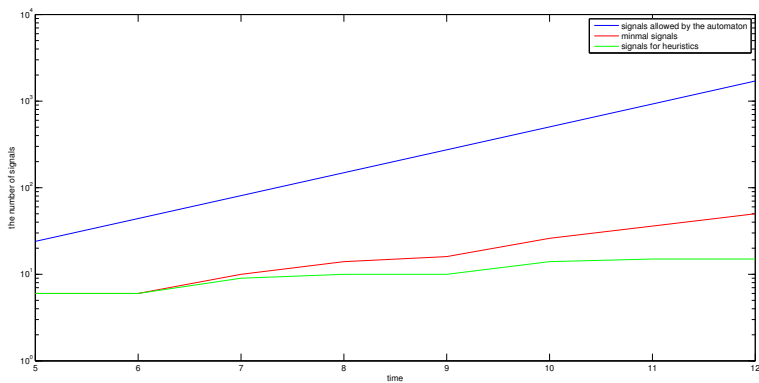


Figure : Automaton: no more than two consecutive dropouts

Examples: Computation time

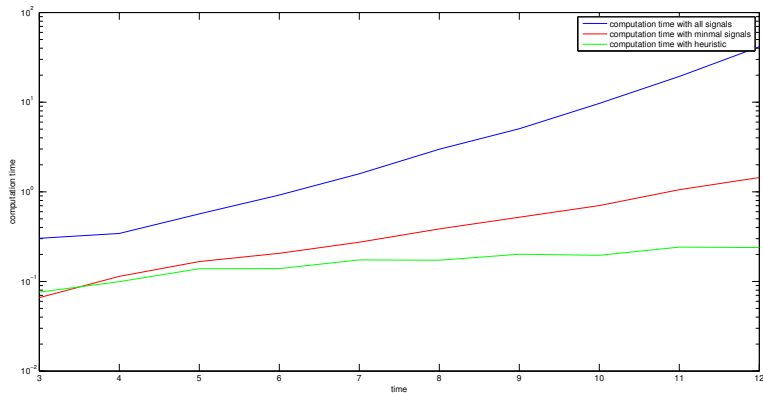


Figure : Automaton: no more than one consecutive dropouts

Examples: Computation time

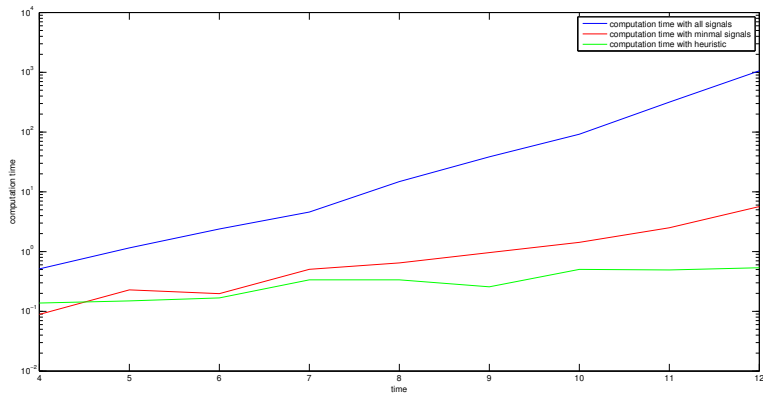


Figure : Automaton: no more than two consecutive dropouts

Table : performance of the heuristic for $T = 12$ (relative percentage error)

(No. of dropouts, dimension, no. of inputs)	no. of samples	zero error	error $\leq 20\%$	21% \leq error $\leq 60\%$	61% \leq error $\leq 90\%$	error $> 90\%$
(1,5,3)	2903	2776	2885	13	4	1
(1,10,4)	2822	2750	2810	10	2	0
(2,5,2)	985	893	938	13	5	29
(2,10,5)	2260	2201	2248	6	0	6
(3,5,3)	2320	2320	2320	0	0	0
(3,8,3)	2878	2875	2876	1	1	0

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Example

$$A = \begin{bmatrix} -1.6747 & -0.4747 & -0.1930 & -0.0423 & -0.3439 \\ -0.4747 & -0.0990 & 1.5974 & -0.1492 & -0.2069 \\ -0.1930 & 1.5974 & -0.3087 & -0.1067 & 0.0434 \\ -0.0423 & -0.1492 & -0.1067 & -1.8911 & 0.5842 \\ -0.3439 & -0.2069 & 0.0434 & 0.5842 & 1.3735 \end{bmatrix}, B = \begin{bmatrix} 0.4497 & 0.2410 \\ 0.8504 & -0.5313 \\ 0.6386 & -0.3696 \\ -0.0176 & 2.6961 \\ -0.1881 & -1.1107 \end{bmatrix}.$$

No more than two successive dropouts are allowed.

Actual minimum eigenvalue of the Gramian = 0.0088, heuristic value = 0.4332.

Corresponding signals:

Actual: $\{0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0\}$,

Approximate: $\{0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0\}$.

Theorem

Suppose (1) is controllable and A has at least one eigenvalue inside the unit circle.

- 1 For any two time instants $t_1 \leq t_2$, $\lambda(t_1) \leq \lambda(t_2)$. Furthermore, $\underline{\lambda}(t) = \lambda(t) \leq \lambda$ and $\lim_{t \rightarrow \infty} \underline{\lambda}(t) = \lambda$.
- 2 Suppose

$$V^{-1}AV = \bar{A} = \begin{bmatrix} A_1 & 0 \\ 0 & A_2 \end{bmatrix}, \bar{B} = V^{-1}B = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix},$$

$$\bar{W}_{\sigma^*(t)} = V^{-1}W_{\sigma^*(t)}V^{-T} \text{ and } \hat{W}_{\sigma^*(t)} = \begin{bmatrix} I & 0 \end{bmatrix}^T \bar{W}_{\sigma^*(t)} \begin{bmatrix} I \\ 0 \end{bmatrix}$$

where A_1 has all eigenvalue strictly inside the unit circle and A_2 has eigenvalues on or outside the unit circle. Then,

$$\lambda \leq \hat{\lambda}(t) \tag{16}$$

$$\text{where } \hat{\lambda}(t) = \|V\|_2^2 \lambda_{\min}(\hat{W}_{\sigma^*(t)} - W_t(A_1, B_1) + W(A_1, B_1)). \tag{17}$$

Corollary

Consider a system of the form (1) which is controllable and stable. Suppose

$$\bar{\lambda}(t) = \lambda_{\min}(W_{\bar{\sigma}}). \quad (18)$$

Then,

- 1 $\bar{\lambda}(t) \geq \lambda$.
- 2 $\lim_{t \rightarrow \infty} \bar{\lambda}(t) = \lambda$.

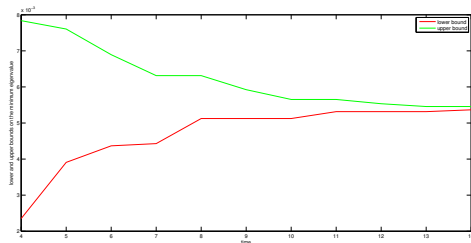










Figure : A stable, Automaton: no more than two consecutive dropouts

- Given a finite amount of energy/average energy, we show that the **reachability problem** is **decidable**.

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- We provide a **heuristic** for the same which reduces the amount of computations further (polynomial time).
- We want to characterize a **trade-off** between the **control energy** and the **cost of the communication network**.
- We wish to consider more general models for decentralized control where **switching signals** associated with **different inputs** are governed by **different automata**.

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THANK YOU.