Supervisory Control of Infinite State Systems under Partial Observation

B. Jeannet, G. Kalyon, <u>T. Le Gall</u>, H. Marchand, T. Massart

Séminaire CIRM - LMeASI

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Outline



- 2 State Avoidance Control Problem
- 3 Control of Systems under Partial Observation
- Decentralized Control Problem
- 5 Distributed Control Problem



State Avoidance Control Problem Control of Systems under Partial Observation Decentralized Control Problem Distributed Control Problem Conclusion

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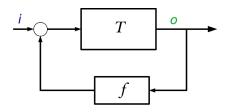


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Once upon a Time in Supervisory Control

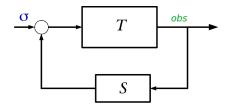


Automation and Control Theory

- Feedback loop to control an electronic device ${\cal T}$
- An input signal i, a control function f, a specification for the output o
- Question : what should be f so that the output signal matches the specification ?

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Once upon a Time in Supervisory Control (2)



Computer science Theory [Ramadge & Wonham 86]

- Discrete event system ${\mathcal T}$
- \bullet The supervisor ${\cal S}$ observes ${\cal T}$ and may disable some events
- Question : what should be ${\cal S}$ so that ${\cal T}||{\cal S}$ satisfies the specification ?

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Formal Definition of the Control Problem

Formalism : Regular languages

- System : $\mathcal{T} = \langle Q, \Sigma, q_0, Q_f,
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 angle$, language M
- Partition : $\Sigma = \Sigma_c \cup \Sigma_{uc}$
- Specification : langage K

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Solution (for prefix-closed languages)

• L is controllable with respect to M if :

 $L.\Sigma_{\mathit{uc}}\cap M\subseteq L$

- The is a unique supremal controllable language $\widehat{L} \subseteq K$
- Supervision : forbid every events not enable by $\widehat{\boldsymbol{L}}$

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Why does he talk about that?

What about static analysis? Abstract interpretation? I want my money back!

Principles of Abstract Interpretation

Static Analysis and Abstract Interpretation

- Static Analysis : method to know what a program can do without running it
- Relies on a fixpoint computation in a lattice (generally $2^{\mathcal{D}_V}$)
- Abstract Interpretation : method to solve this fixpoint computation using approximation

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Outline of the method

- Concrete lattice $2^{\mathcal{D}_V} \xrightarrow{\gamma} \Lambda$ Abstract lattice
- We transpose the computation into the abstract lattice
- $\bullet\,$ A widening operator ∇ ensures the convergence of this computation
- The obtained solution is an overapproximation of the least fixpoint

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From Supervisory Control to Abstract interpretation

Supremal controllable language = fixpoint computation

- Controllable = (post-) fixpoint
- State-based approach : if K is a set of "good" states, we want the greatest fixpoint of X → Post_{uc}(X) ∩ K
- Supervision : disable all transitions leading out this set of states

What about Abstract Interpretation?

- When reachability is undecidable, we may obtain a valid supervisor
- Over-approximation of a least fixpoint
- Safe control, but possible loss of permissivity

Outline



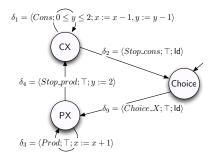
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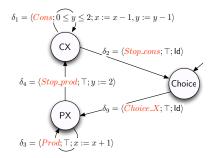
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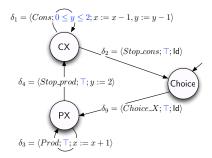




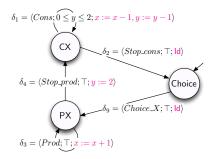
- Numerical variables, domain \mathcal{D}_V
- Symbolic Transitions $\delta = \langle \sigma, G, A \rangle$
- Semantics : infinite Labelled Transition System (LTS)



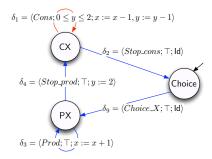
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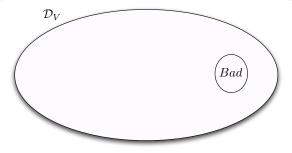
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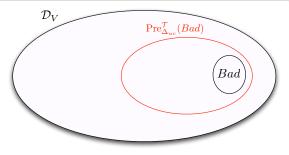


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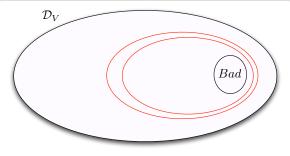
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- Symbolic Transitions $\delta = \langle \sigma, G, A \rangle$
- Semantics : infinite Labelled Transition System (LTS)
- Transitions $\Delta = \Delta_c \uplus \Delta_{uc}$
- Objective : avoid $Bad \subseteq \mathcal{D}_V$



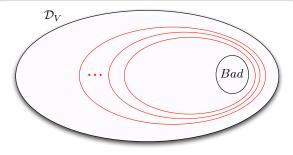


Fixpoint computation

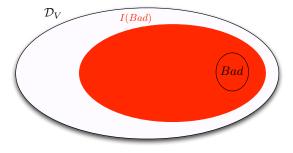
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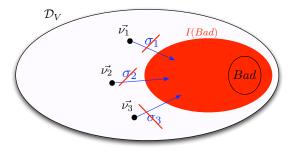
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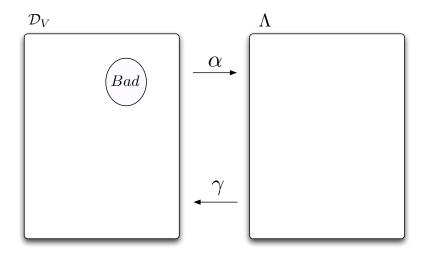
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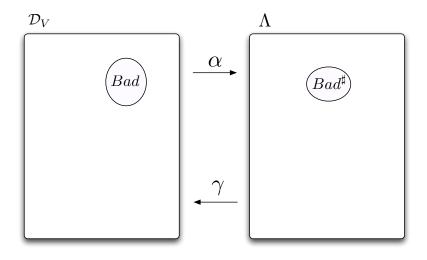


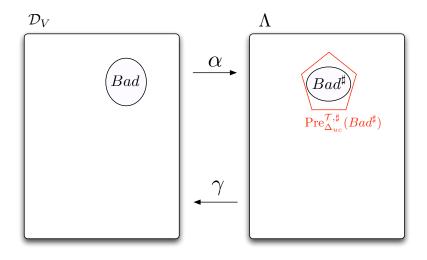
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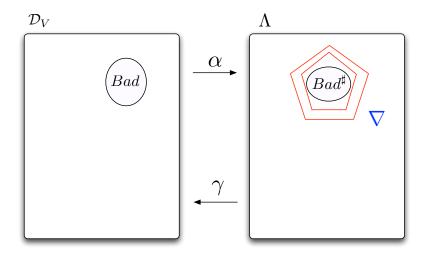


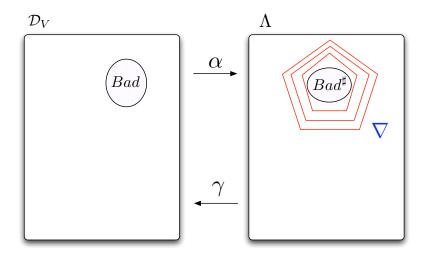
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- Supervisor : forbid all transitions leading to *I*(*Bad*)

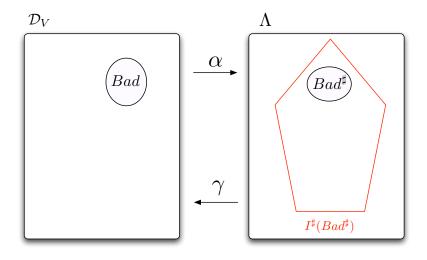


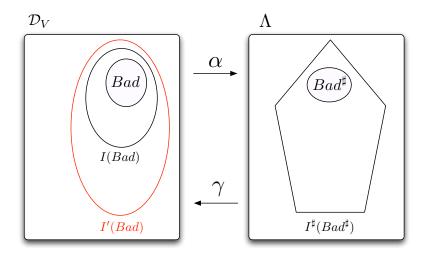


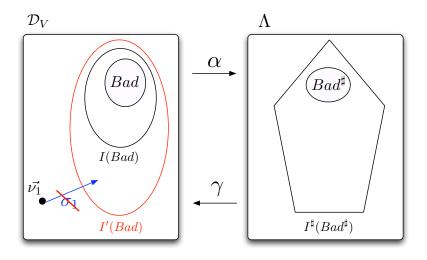












Permissiveness and non-blocking control

Permissiveness

- Quality of control : how permissive the supervisor is
- More precise fixpoint computation \Rightarrow better supervisor
- Choice of the abstract lattice, of the fixpoint computation strategy, ...

Permissiveness and non-blocking control

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Permissiveness and non-blocking control

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Deadlock-free case

- Deadlock-free : at least one event is enable in every state
- Non-blocking : always able to reach a given objective
- We "solved" the deadlock-free case, underapproximations are needed so solve the non-blocking problem

(Partial) Conclusion

Advantages of the approach

- For the "supervisory control" community :
 - Rigorous fixpoint computation method
 - Termination (enven when the problem is undecidable)
- For the "static analysis" community :
 - A new playground !
 - Some problems that do not occur in software verification (e.g. non-blocking)



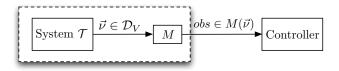
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Systems under Partial Observation



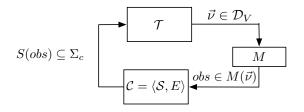
Partial observation

- Modeled by an observer $\langle \mathcal{D}_{Obs}, M \rangle$
- The observation space \mathcal{D}_{Obs} can be infinite
- $M: \mathcal{D}_V \mapsto 2^{\mathcal{D}_{Obs}}$ is a mask

Example of observers

- Hidden variables : $M(\langle \mathsf{PX}, 10, 15 \rangle) = \langle \mathsf{PX}, 10 \rangle$
- Undistinguishable locations : $M(\langle \mathsf{PX}, 10, 15 \rangle) = M(\langle \mathsf{Choice}, 10, 15 \rangle)$
- . . .

Memoryless Controllers

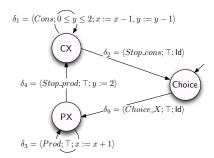


Formalization of the memoryless controller

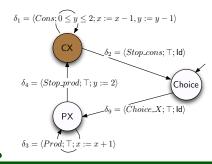
- Restricts the behavior of the system according to the observation
- The controller is a pair $\mathcal{C} = \langle \mathcal{S}, E \rangle$:
 - O The supervisory function S : D_{Obs} → 2^{Σ_c} gives the sets S(obs) of controllable actions to forbid in obs ⇒ memoryless controller

2 $E \subseteq \mathcal{D}_V$ restricts the set of initial states

Example

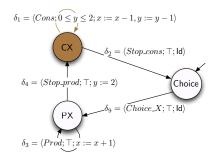


- A state is a tuple $\langle \ell, x, y \rangle \subseteq Loc \times \mathbb{N} \times \mathbb{N}$
- The mask *M* is defined as follows for each state $\vec{\nu} = \langle \ell, x, y \rangle$:
 - if $x \notin [10, 20]$, then $\vec{\nu}$ is perfectly observed
 - otherwise, $\vec{\nu}$ is undistinguishable from $\{\langle \ell, x_1, y \rangle \, | \, x_1 \in [10, 20]\}$
- $Bad = \{ \langle \mathsf{CX}, x, y \rangle | (x \le 10) \land (0 \le y \le 2) \}$

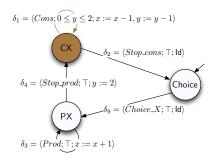


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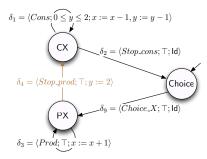


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- $\cup \{\langle \mathsf{CX}, x, y \rangle | (x \le 11) \land (1 \le y \le 2)\} \cup \{\langle \mathsf{CX}, x, y \rangle | (x \le 12) \land (y = 2)\}$

Example

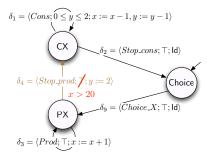


As a reminder : $I(Bad) = \{ \langle \mathsf{CX}, x, y \rangle | (x \le 10) \land (0 \le y \le 2) \} \cup \{ \langle \mathsf{CX}, x, y \rangle | (x \le 11) \land (1 \le y \le 2) \} \cup \{ \langle \mathsf{CX}, x, y \rangle | (x \le 12) \land (y = 2) \}.$

Control function

Stop_prod forbidden in $M^{-1}(M(\{\langle \mathsf{PX}, x, y \rangle | x \le 12\})) = \{\langle \mathsf{PX}, x, y \rangle | x \le 20\}$

Example



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Beyond Memoryless Controller

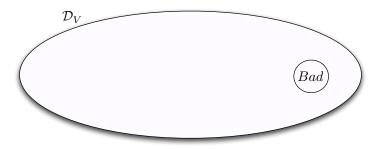
Memory improves the control decision

- Memoryless controller must take their decision on a single observation
- What about states that have the same observation, but can be distinguished by the past execution?

Improvement of the controller

- *k*-memory controllers
- Online controllers

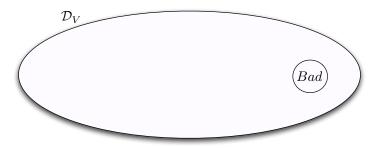
Online Controllers



Online controllers

• The controller maintains an estimate of the current state of ${\mathcal T}$ to define its control policy

Online Controllers



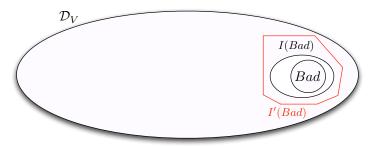
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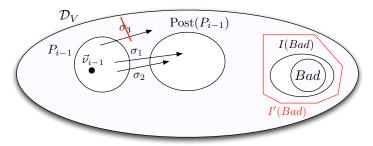
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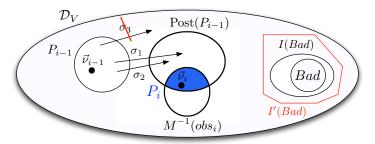
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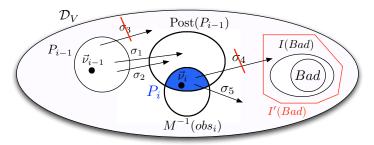
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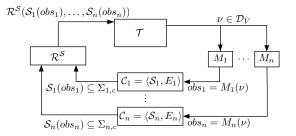
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Outline



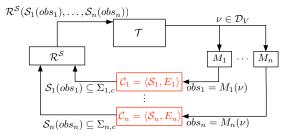
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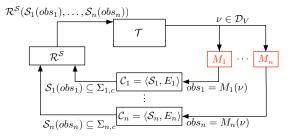


Decentralized Approach

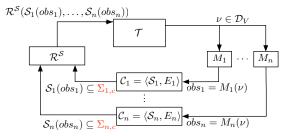
• The decentralized approach is more suitable for the control of distributed systems with synchronous communications



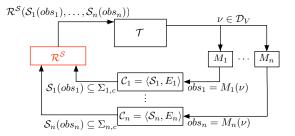
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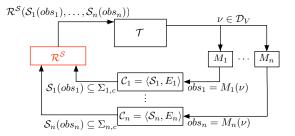
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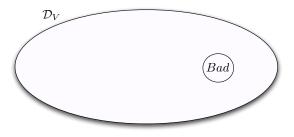


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- A synchronization mechanism, called fusion rule, defines the global control to be applied to the system from the control decisions of the controllers C_i : an action is forbidden if each controller controlling this action proposes to forbid it

Decentralized Controller(2)

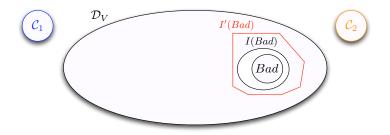


Basic decentralized Problem

To synthesize *n* valid and non-trivial controllers i.e., *n* controllers that

- prevent from reaching Bad
- do not reduce the behavior of the controlled system to the empty set

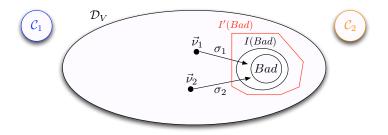
Decentralized Controller(2)



Control Policy

• Each controller C_i forbids the actions that it controls and that lead to I'(Bad)

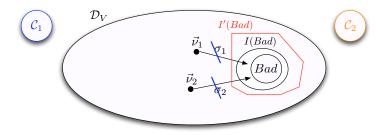
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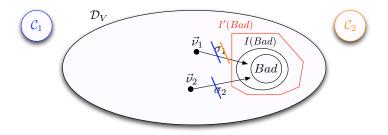
Decentralized Controller(2)



Control Policy

 Each controller C_i forbids the actions that it controls and that lead to I'(Bad) C₁ forbids σ₁ in ν₁ C₁ forbids σ₂ in ν₂

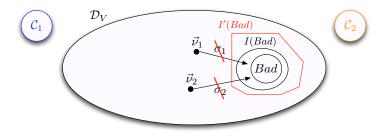
Decentralized Controller(2)



Control Policy

- Each controller C_i forbids the actions that it controls and that lead to I'(Bad) C_1 forbids σ_1 in $\vec{\nu}_1$ C_2 forbids σ_1 in $\vec{\nu}_1$
 - C_1 forbids σ_2 in $\vec{\nu}_2$ C_2 allows σ_2 in $\vec{\nu}_2$ (uncontrollable action)

Decentralized Controller(2)



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 C_2 allows σ_2 in $\vec{\nu}_2$ (uncontrollable action)

 $\mathcal{R}^{\mathcal{S}}$ forbids σ_1 in $\vec{\nu}_1$ \mathcal{R}^{S} forbids σ_{2} in $\vec{\nu}_{2}$ (because \mathcal{C}_{2} does not control σ_{2})

Model and Control Objective State Estimates Solution of the Control Problem

Outline



- 2 State Avoidance Control Problem
- Control of Systems under Partial Observation
- 4 Decentralized Control Problem
- Distributed Control Problem



Model and Control Objective State Estimates Solution of the Control Problem

Distributed Systems with Asynchronous Communications

Motivation

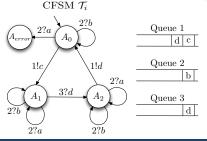
- The decentralized and modular approaches cannot be used for the control of distributed systems with asynchronous communications
- The communications between the subsystems are not instantaneous

Distributed Approach

- We propose a distributed approach that takes into account the asynchronous nature of communications
- We consider distributed systems T composed of several subsystems T_i communicating through reliable unbounded FIFO channels
- T_i is modeled by a communicating finite state machine (CFSM)

Model and Control Objective State Estimates Solution of the Control Problem

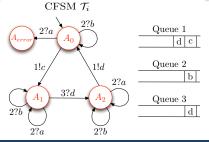
System to be Controlled



- Finite set L_i of locations
- Reliable and unbounded FIFO channels
- Finite set Σ_i of actions. An action σ is :
 - either an output Q!m
 - or an input Q?m
- Finite set of transitions $\langle \ell_i, \sigma, \ell'_i \rangle$
- Semantics = infinite Labeled Transition System (LTS)

Model and Control Objective State Estimates Solution of the Control Problem

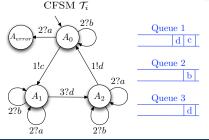
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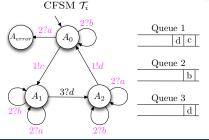
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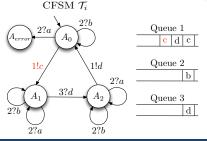
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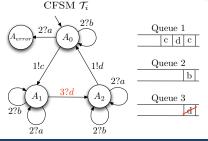
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Model and Control Objective State Estimates Solution of the Control Problem

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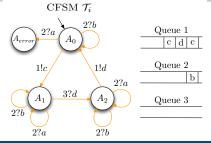


Communicating Finite State Machine \mathcal{T}_i

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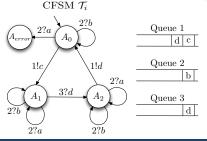


Communicating Finite State Machine T_i

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Model and Control Objective State Estimates Solution of the Control Problem

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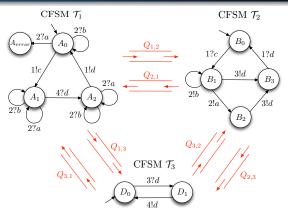


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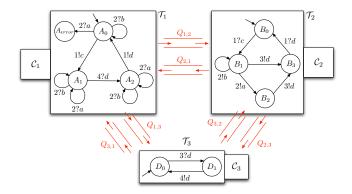


Communication Architecture

There two queues between each pair of subsystems T_i and T_j :

- $Q_{i,j}$: \mathcal{T}_i writes on this queue and \mathcal{T}_j reads the sent messages
- $Q_{j,i}$: \mathcal{T}_i reads on this queue and \mathcal{T}_j writes on it

Means of Observation and Control



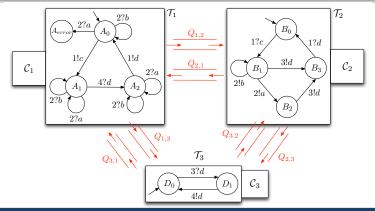
Observation

- Each subsystem T_i is controlled by a controller C_i
- C_i only observes the current state of T_i

Control Mechanism

• For each \mathcal{T}_i , the set $\Sigma_i = \Sigma_{i,c} \uplus \Sigma_{i,uc}$

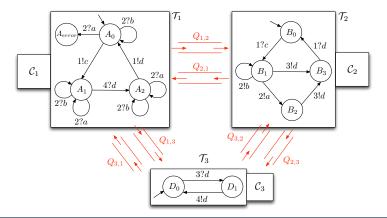
Formalization of the Controllers



Communication

- The controllers C_i communicate with each other to have a better knowledge of the system T
- They communicate by adding information to the messages normally exchanged by the subsystems
- Each controller C_i uses the exchanged information and the information received from T_i to compute an estimate of the current state of the distributed system T

Formalization of the Controllers



Formalization of the controllers $C_i = \langle S_i, E_i \rangle$

- S_i defines, for each set P of states of T, a set S_i(P) of controllable actions that T_i cannot execute when P is the estimate of the current state of T computed by C_i.
- **2** E_i restricts the set of initial states of \mathcal{T} .

Model and Control Objective State Estimates Solution of the Control Problem

Problem

Distributed Problem

To synthesize valid and non-trivial controllers i.e., controllers that

- prevent from reaching Bad
- do not reduce the behavior of the controlled system to the empty set

Model and Control Objective State Estimates Solution of the Control Problem

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Property

The distributed problem is undecidable

Model and Control Objective State Estimates Solution of the Control Problem

Problem

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To synthesize valid and non-trivial controllers i.e., controllers that

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Property

The distributed problem is undecidable

Our approach

- Online controllers
- States estimates computed using abstract interpretation
- Communication (without explicit synchronisation) between controllers

Model and Control Objective State Estimates Solution of the Control Problem

Algorithm for the Distributed Problem

Outline

Synthesize *n* controllers that compute, during the execution of the distributed system \mathcal{T} , estimates of the current state of \mathcal{T} and define their control policy from their state estimates. Since the system is asynchronous, a state estimates for controller C_i must "guess" the possible future behaviour of the other subsystems \mathcal{T}_i , $j \neq i$.

Model and Control Objective State Estimates Solution of the Control Problem

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Algorithm computing the state estimates

- C_i maintains a estimate CS_i of the current state of T
- The controllers exchange their state estimate by adding these information to the messages normally exchanged by the subsystems
- For each transition fired by T_i , the controller C_i updates CS_i from the information received from T_i and the other controllers

Model and Control Objective State Estimates Solution of the Control Problem

Computation of the state estimates

Update of CS_i after the output transition $\delta = \langle \ell_i, Q_{i,j} | m, \ell'_i \rangle$

When \mathcal{T}_i wants to send a message m to \mathcal{T}_j on the queue $Q_{i,j}$:

- CS_i is sent to C_j with m
- CS_i is updated :
 - Post $_{\delta}^{\mathcal{T}}(CS_i)$: takes into account the execution of δ
 - Reachable $\mathcal{T}_{\Delta \setminus \Delta_i}(\mathsf{Post}^{\mathcal{T}}_{\delta}(\mathsf{CS}_i))$:

Model and Control Objective State Estimates Solution of the Control Problem

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takes into account the fact that \mathcal{T}_k ($\forall k \neq i$) continues its execution

Update of CS_i after the input transition $\delta = \langle \ell_i, Q_{j,i}?m, \ell'_i \rangle$

- C_i tries to refine its knowledge of the system using the state estimate CS_j
- Roughly an intersection, followed by $\mathsf{Post}_\delta : CS_i := \mathsf{Post}_\delta(CS_i \cap CS_j)$

Model and Control Objective State Estimates Solution of the Control Problem

Computation of the state estimates

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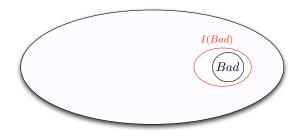
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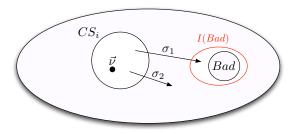
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- *CS_j* may be out-of-date, we may have to update it first, or ignore it. We use vector clocks to detect out-of-date messages



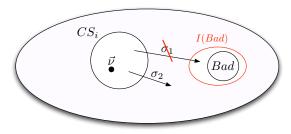
Semi-algorithm

• Offline part : Computation of the set I(Bad) of states of \mathcal{T} leading uncontrollably to Bad



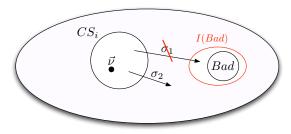
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Effective Algorithm by Means of Abstract Interpretation

- Abstract lattice : regular languages is used to abstract the content of the FIFO channels
- Widening operator : works on regular languages

Model and Control Objective State Estimates Solution of the Control Problem

Abstract Lattices for FIFO Systems

Regular languages as an abstract lattice

- Concrete lattice : 2^{Σ^*} , abstract lattice $\operatorname{Reg}(\Sigma)$
- Canonical representation : minimal deterministic finite automomata
- When there are several FIFO channels : QDD representation
- Representation framework (no Galois connection)

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Widening operator

- Idea : quotient the automaton by an equivalence relation
- (colored) k-bounded bisimulation
- Convergence : number of equivalence classes is bounded (depends on k and $|\Sigma|$)

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Remark : we may work with an infinite alphabet of messages

Control of Systems under Partial Observation Decentralized Control Problem Conclusion

Outline





Conclusion

Contributions

- Abstract interpretation is an efficient tool for controller synthesis
- The "real" problems : partial observation, interaction between controllers

Open questions

- Non-blocking properties : is it possible to use under-approximations ?
- Is it interesting for LMeASI?