

# Specification and Verification of Diagnosis and Predictability in Probabilistic Systems

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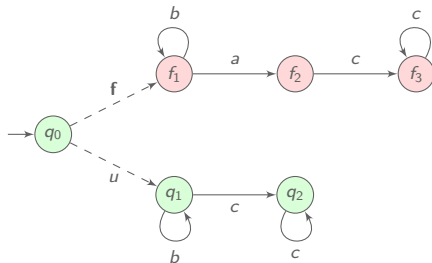


# Diagnosis Framework

*LTS*: Labelled transition system.

*Diagnoser*: must tell whether a fault  $f$  occurred, based on observations.

*Convergence hypothesis*: no infinite sequence of unobservable events.

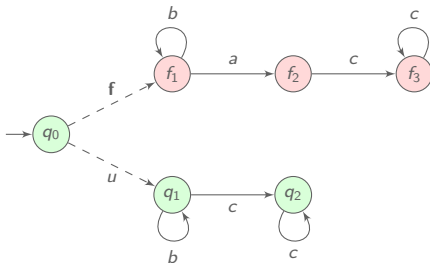


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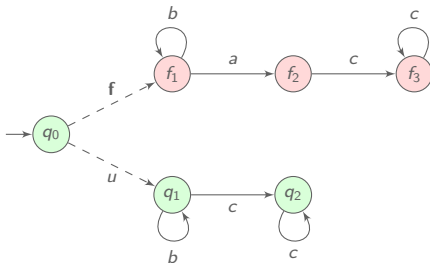
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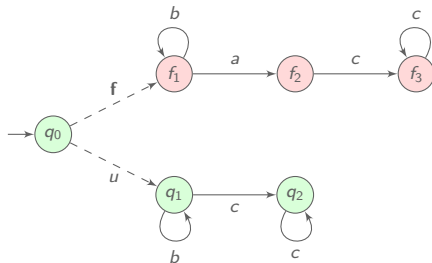
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**?**  $b$  is ambiguous as  $\mathcal{P}^{-1}(b) = \{q_0 \xrightarrow{f} f_1 \xrightarrow{b} f_1, q_0 \xrightarrow{u} q_1 \xrightarrow{b} q_1\}$ .

# Diagnosis Problems

Diagnoser requirements:

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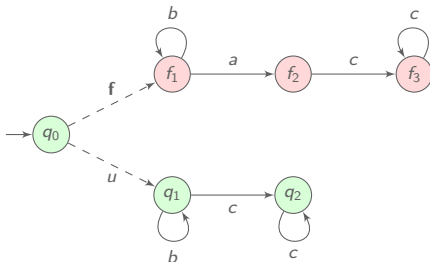
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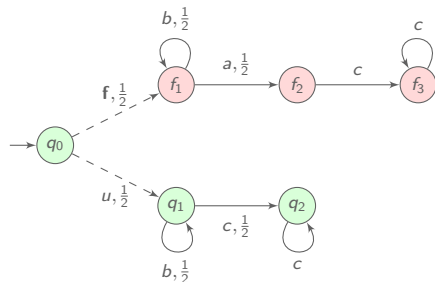
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A sound but not reactive diagnoser : claiming a fault when  $a$  occurs.



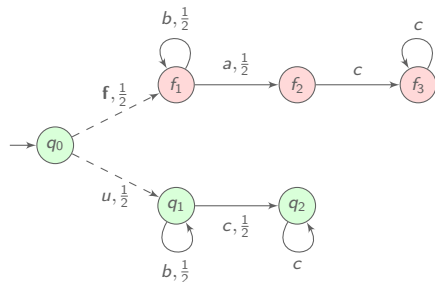
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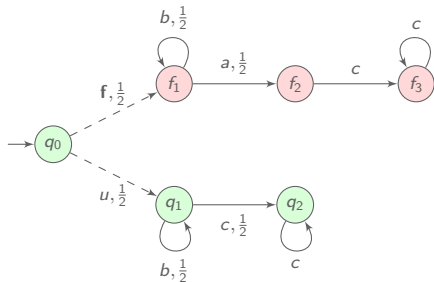


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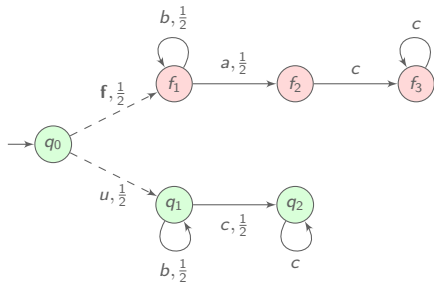
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How to adapt soundness and reactivity?

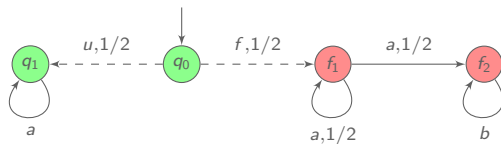
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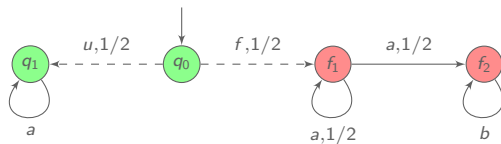
# Outline

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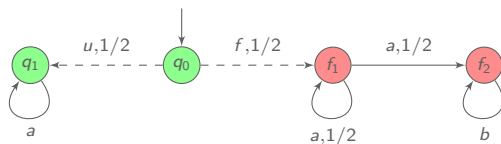
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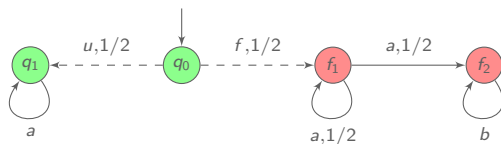
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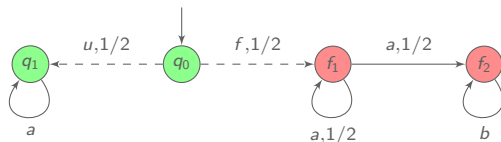


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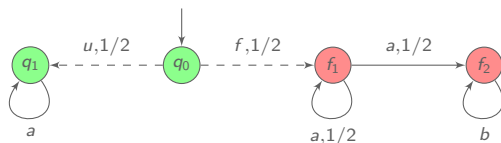
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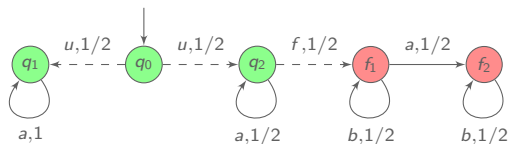
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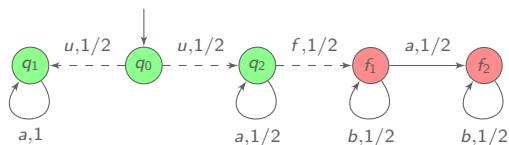
## Two reactivity specifications:

- ▶ Detect a fault almost surely.
- ▶ Detect if a run is faulty or correct almost surely.

# Infinite sequences or their finite prefixes?

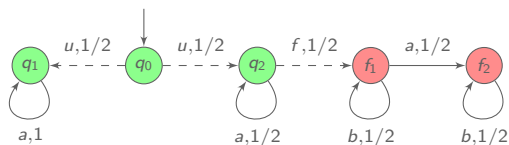


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$a^n$  is ambiguous and  
 $\mathbb{P}(q_0 u (q_1 a)^n) = \frac{1}{2}$ .

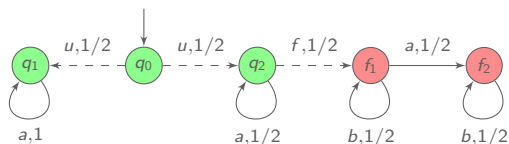
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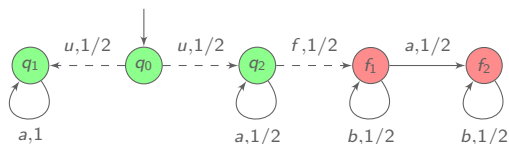
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- ▶ The probability of ambiguous sequences goes to 0 when the length goes to infinity.
- ▶ An infinite sequence is almost surely non ambiguous.

# Four specifications of diagnosis

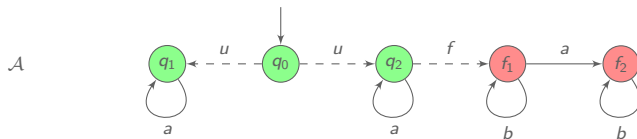
| Diagnosability     | All runs              |                  | Faulty runs             |
|--------------------|-----------------------|------------------|-------------------------|
| Finite prefixes    | FA                    | $\Rightarrow$    | FF                      |
|                    |                       | $\not\Leftarrow$ |                         |
| Infinite sequences | $\Downarrow \Uparrow$ |                  | $\Downarrow \Uparrow^*$ |
|                    | IA                    | $\Rightarrow$    | IF                      |
|                    |                       | $\not\Leftarrow$ |                         |

\* assuming finitely-branching models

# Outline

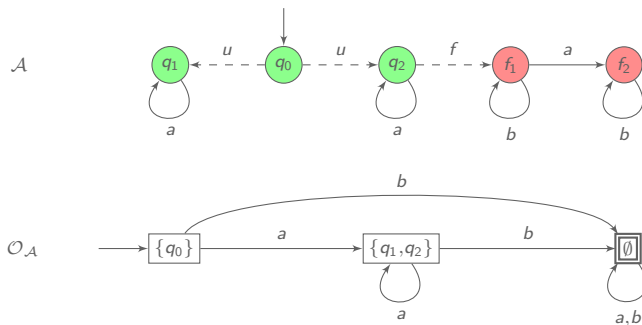
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- Specification of IF: 1) Infinite sequences  
2) Fault diagnosis



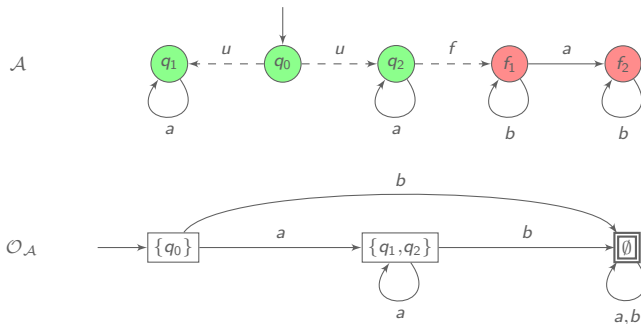
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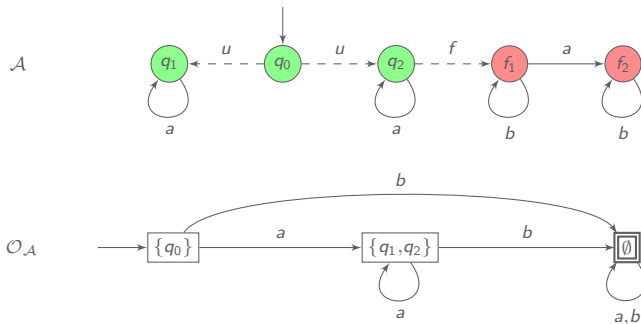


$\mathcal{A}$  is not IF-diagnosable  
iff

there exists a state  $(q, U)$  in a BSCC of  $\mathcal{A} \times \mathcal{O}_{\mathcal{A}}$  with  $q$  faulty and  $U \neq \emptyset$ .

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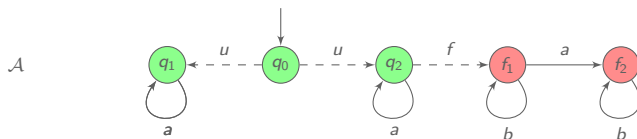
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If  $\mathcal{A}$  is a IF-diagnosable pLTS with  $n$  correct states  
then  $\mathcal{O}_{\mathcal{A}}$  is a IF-diagnoser of  $\mathcal{A}$  with at most  $2^n$  states.

# IA-diagnosability

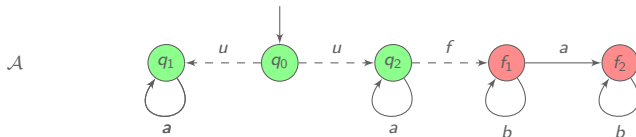
- Specification of IA: 1) Infinite sequences  
2) All sequences disambiguation



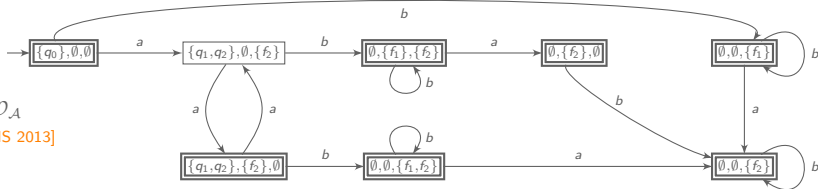


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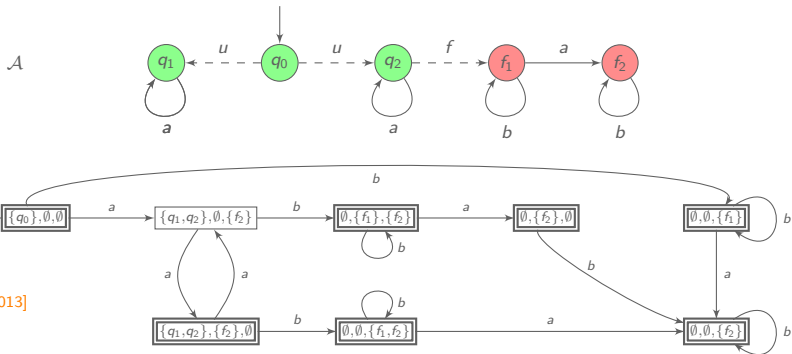


$\mathcal{O}_{\mathcal{A}}$   
[HHMS 2013]



# IA-diagnosability

- Specification of IA: 1) Infinite sequences  
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$\mathcal{A}$  is not IA-diagnosable iff  
there exists a BSCC of  $\mathcal{A} \times \mathcal{O}_{\mathcal{A}}$  where every state  $(q, U, V, W)$  satisfies  
 $q$  faulty and  $U \neq \emptyset$  or  $q$  correct and  $W \neq \emptyset$ .

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## Sketch of proof

- ▶ relies on the characterisation on  $\mathcal{A} \times \mathcal{O}_{\mathcal{A}}$
- ▶ avoids building the product
- ▶ uses Savitch's theorem for appropriate guesses

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# Diagnosability is PSPACE-hard

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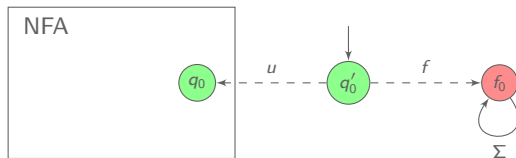
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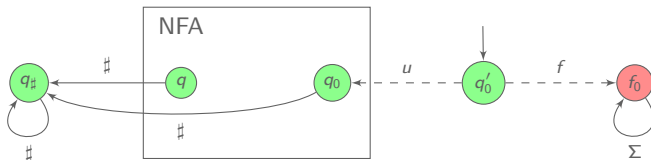
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**Erroneous** adaptation to stochastic systems in [Chen/Kumar-tase13].

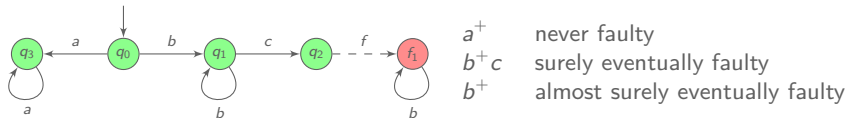
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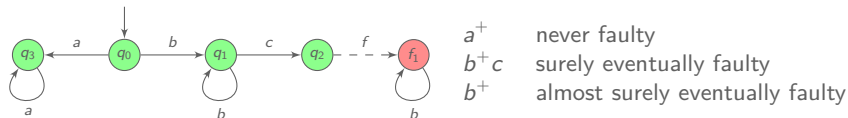
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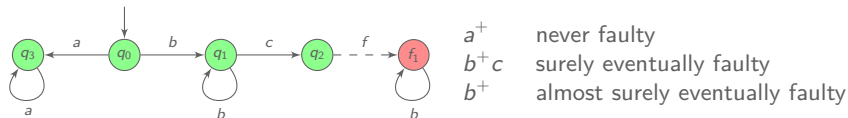
## Two notions of soundness:

- ▶ sure: if a fault is claimed, a fault will occur
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**Reactivity:** a fault is detected at least  $k$  steps before occurrence.

# Sure and almost-sure $k$ -predictability

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- ▶ surely 0-predictable
- ▶ almost surely 1-predictable
- ▶ not 2-predictable

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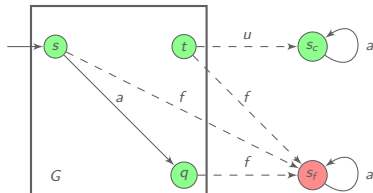
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- ▶ **Upper bound:** graph characterisation
- ▶ **Lower bound:** reduction of reachability in acyclic graphs



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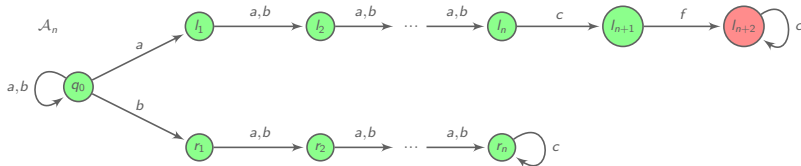
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$\mathcal{A}_n$  has  $2n + 2$  correct states and every (almost) sure 0-predictor of  $\mathcal{A}_n$  needs at least  $2^n$  states.

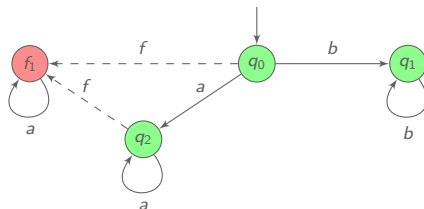




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**Prediagnoser:** detects and foresees faults analysing past and future

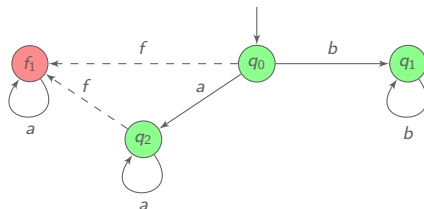


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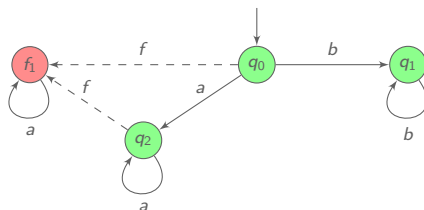
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# Conclusion: Foundation of stochastic diagnosis

## Summary of contributions

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## Future work

- ▶ Approximate diagnosis (relaxing soundness)
- ▶ Other paradigms related to partial observation (detectability, opacity, etc.)
- ▶ Space and time optimisation of observations