

# Predictability of Event Occurrences in Timed Systems

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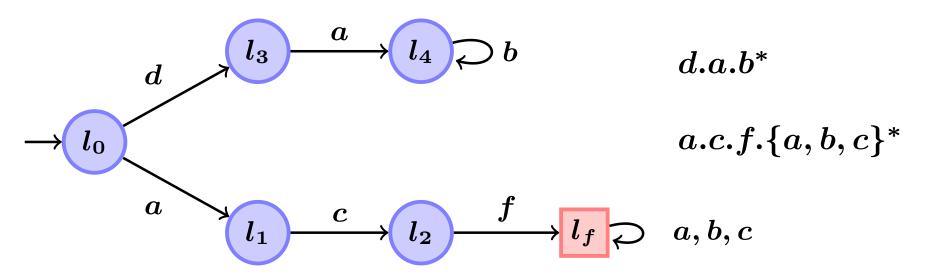
NICTA and UNSW, Australia

NICTA and ANU, Australia

FORMATS 2013, Buenos Aires, August 2013



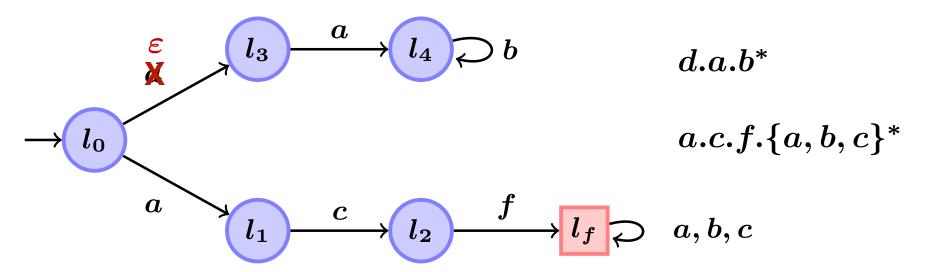




#### System

- generates sequences of events
- model is available

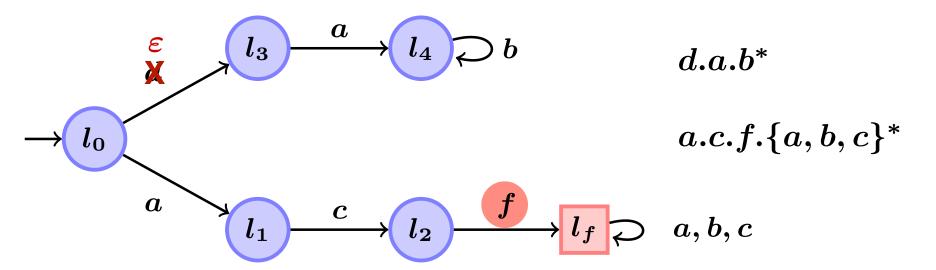




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- model is available
- partially observable



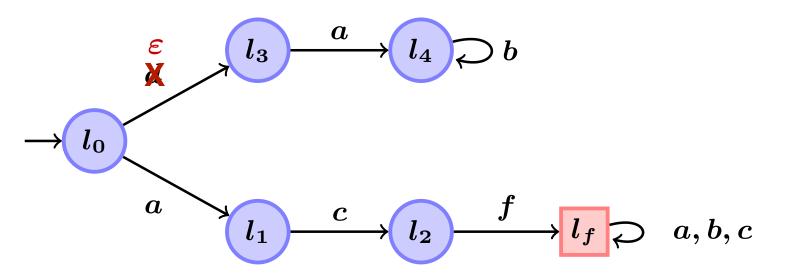


#### System

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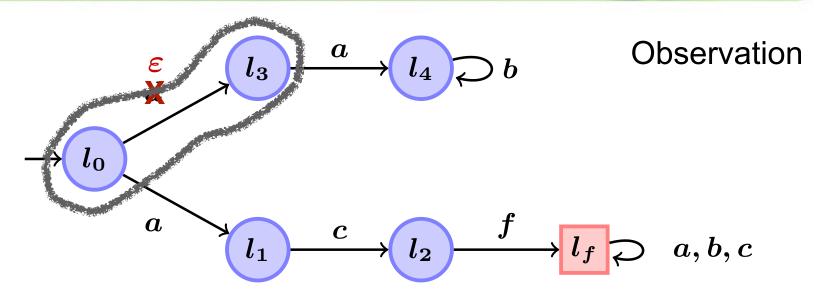
Goal: predict event f





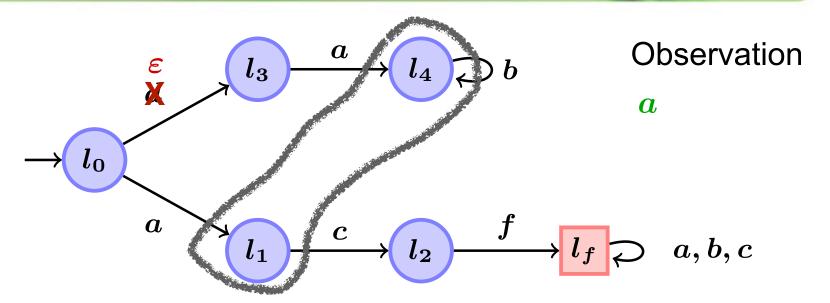
$$\pi(d.a.b.b) = a.b.b$$





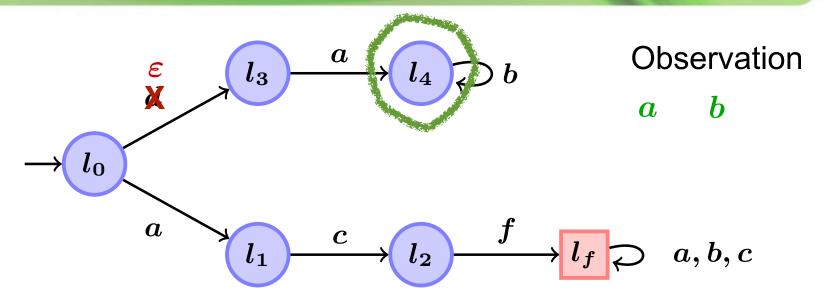
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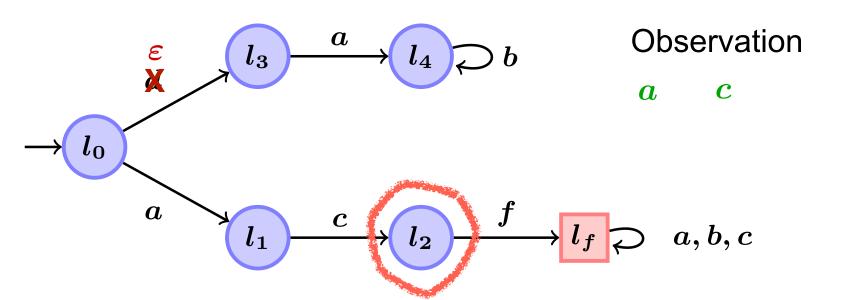
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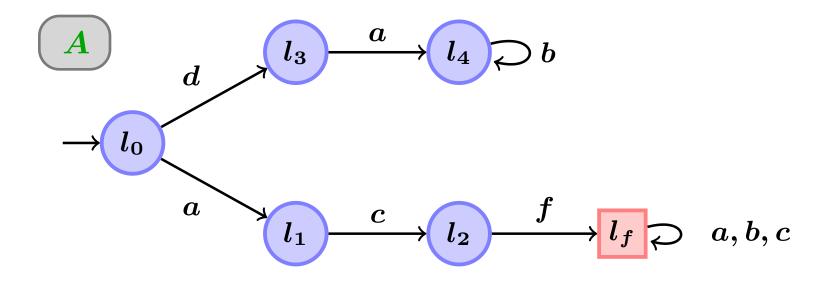
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#### Plan



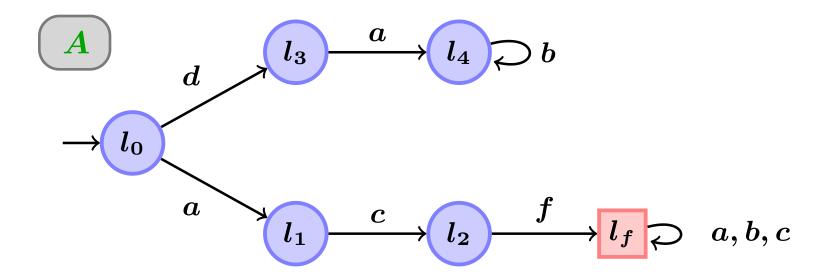
- Formal simple definition of predictability
   extends previous definition by Genc and Lafortune
   language based and valid for timed systems
- 2) Predictability for discrete time systems polynomial time algorithm
- 3) Predictability for timed systems PSPACE-completeness
- 4) Sampling predictability implementable predictors





$$L(A)$$
 = finite or infinite traces of  $A$   $d.a.b^{\omega} \ a.c.f.b.a$ 





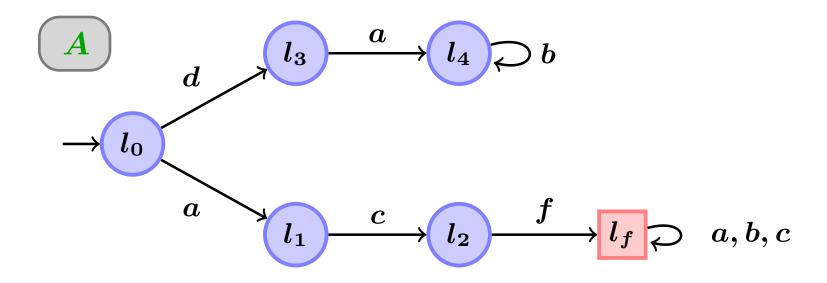
L(A) = finite or infinite traces of A

 $d.a.b^{\omega}$  a.c.f.b.a

 $L^{\omega}_{\neg f}$  = infinite traces of A with no f

 $d.a.b^{\omega}$ 





$$L(A)$$
 = finite or infinite traces of  $A$ 

$$d.a.b^{\omega} \ a.c.f.b.a$$

$$L_{\neg f}^{\omega}$$
 = infinite traces of  $A$  with no  $f$ 

$$d.a.b^{\omega}$$

$$egin{array}{ll} L^{\omega}_{\neg f} &= & \text{infinite traces of } A \text{ with no } f \\ L^{-k}_f &= & \text{finite traces } w \text{ of } A \text{ with no } f \end{array}$$

$$a.c \in L_f^{-0} \ a \in L_f^{-1}$$

such that 
$$egin{cases} w.x.f \in L(A) \ |x| \leq k \end{cases}$$



Partial observation: projection  $\pi(w)$ 

(1) k -predictor

(2) CNS for k predictbility and predictability

(3) boxes beige color with equivalence

(4)

k-predictor: mapping *P* that satisfies:

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k-predictability = existence of a k-predictor predictability =  $\exists k$  such that k-predictable



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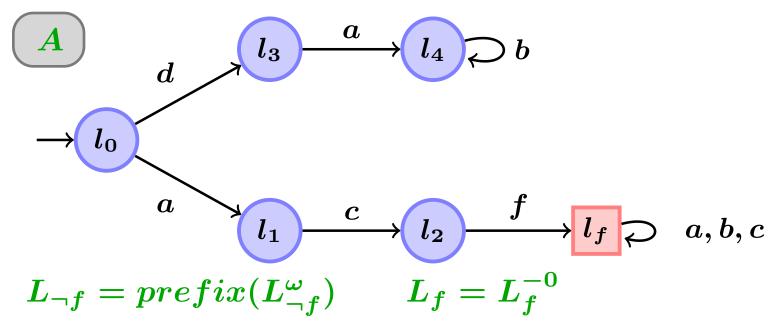
k-predictability = existence of a k-predictor predictability =  $\exists k$  such that k-predictable

$$k$$
-predictability  $\iff \pi(prefix(L^{\omega}_{\lnot f})) \, \cap \, \pi(L^{-k}_f) = arnothing$ 

predictability 
$$\iff$$
 0-predictability

### Genc and Lafortune Predictability





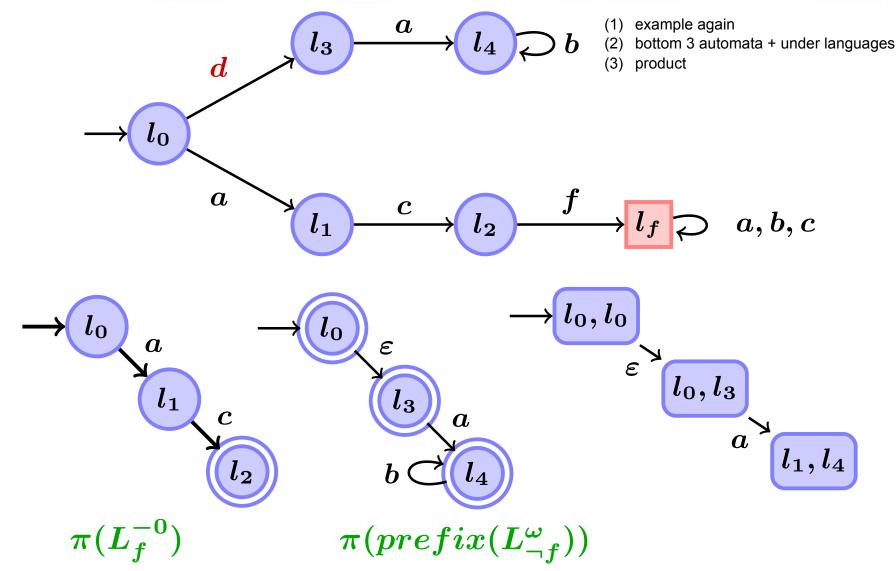
GL-predictability:

 $egin{aligned} &\exists n \in \mathbb{N}, orall w \in L_f, \exists t \in prefix(w) ext{ such that } \mathbf{F}(t) \ &\mathbf{F}(t): orall u \in L_{\lnot f}, orall v = u.x \in \mathcal{L}(A), \ &\pi(u) = \pi(t) \land |v| \geq n \implies |v|_f > 0. \end{aligned}$ 

GL-predictability  $\iff$  0-predictability

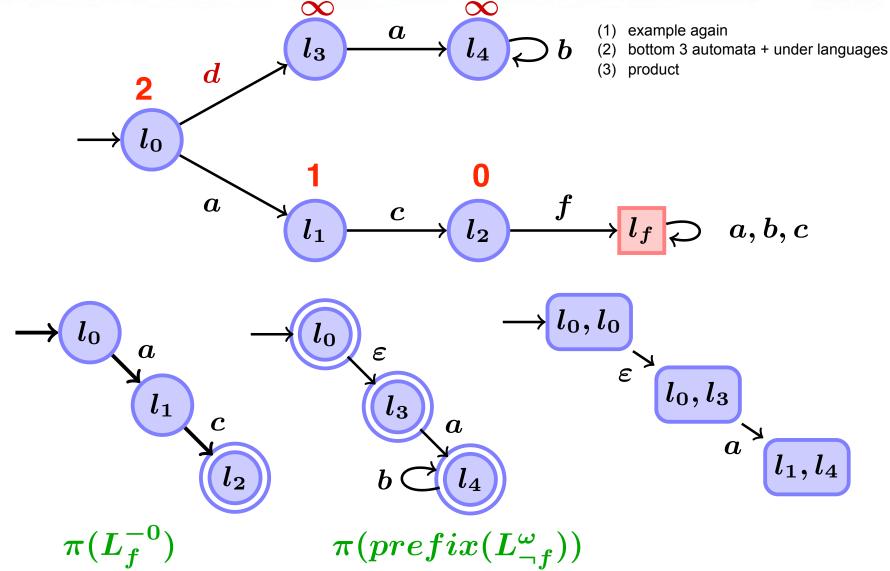
### **Checking k-Predictability**





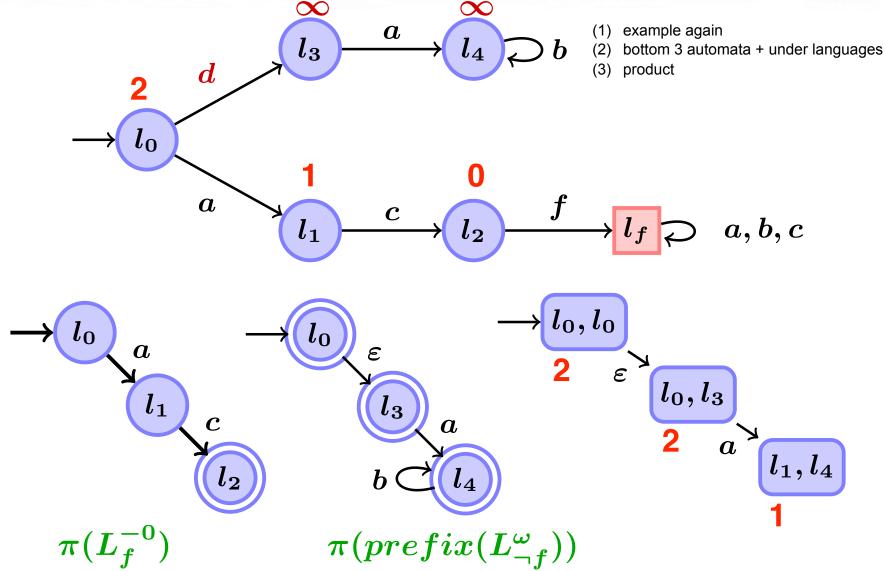
#### Computing the Maximum k





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### Computing the k-Predictor



If k-predictor exists:

$$\begin{cases} \forall w \in \mathit{prefix}(L^{\omega}_{\neg f}), P(\pi(w)) = 0 \\ \forall w \in L^{-k}_f, P(\pi(w)) = 1 \end{cases}$$

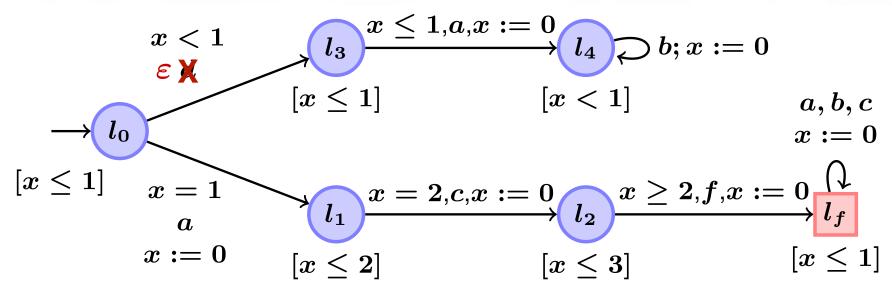
and  $\pi(prefix(L^{\omega}_{\lnot f})) \, \cap \, \pi(L^{-k}_f) = arnothing$ 

Build a deterministic automaton that accepts  $\pi(L_f^{-k})$  or

on-the-fly determinisation

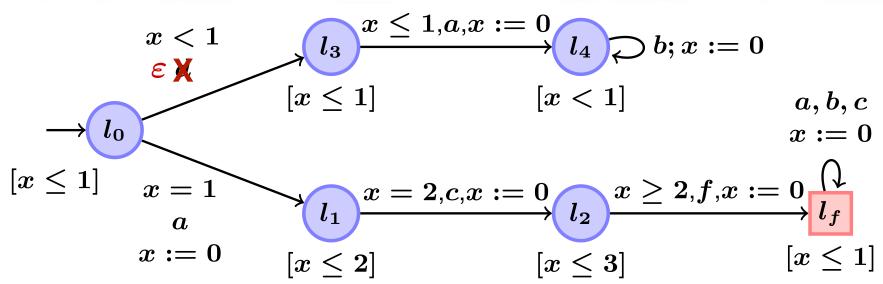
If  $m{A}$  is predictable, there is a finite state predictor Worst-case: exponential size in  $m{A}$ 





$$L(A)$$
 = timed traces of  $A$  
$$1~a~2~c~2.5~f$$
 projection:  $\pi(0.6~d~0.2~a~0.1~b) = (0.6+0.2)~a~0.1~b$ 

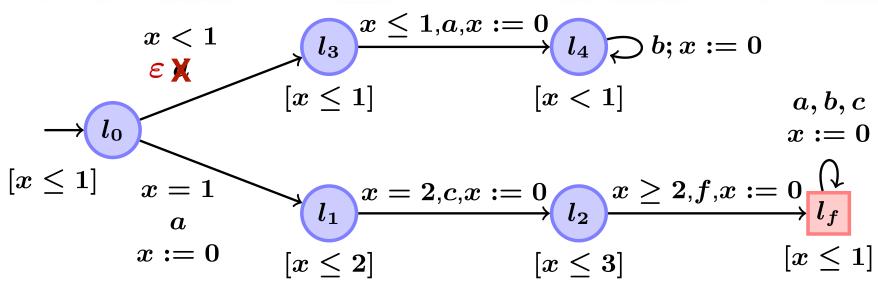




$$L(A)$$
 = timed traces of  $A$  1  $a$  2  $c$  2.5  $f$  projection:  $\pi(0.6\ d\ 0.2\ a\ 0.1\ b) = (0.6+0.2)\ a\ 0.1\ b$   $L^{\omega}_{\neg f}$  = infinite timed traces of  $A$  with no  $f$  0.6  $d$  0.2  $a$  (0.1  $b$ ) $^{\omega}$ 

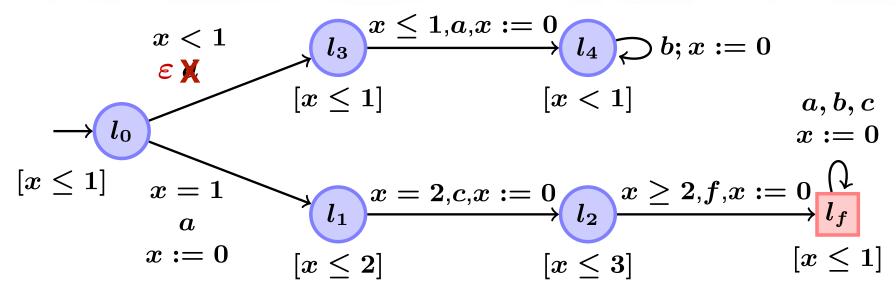


 $1\ a\ 2\ c\ 2\in L_f^{-0}$ 



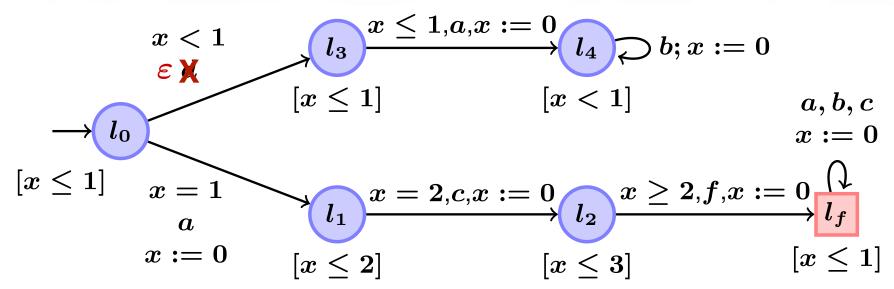
$$L(A) = \text{timed traces of } A \qquad \qquad 1 \ a \ 2 \ c \ 2.5 \ f$$
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$$L^{\omega}_{\neg f} = \text{infinite timed traces of } A \text{ with no } f \qquad 0.6 \ d \ 0.2 \ a \ (0.1 \ b)^{\omega}$$
 
$$L^{-k}_{f} = \text{finite timed traces } w \text{ of } A \text{ with no } f \qquad 1 \ a \ 2 \ c \in L^{-3}_{f}$$
 
$$\text{such that } \begin{cases} w.x.f \in L(A) & 1 \ a \ 2 \ c \in L^{-2}_{f} \\ dur(x) \le k & 1 \ a \ 2 \ c \in L^{-2}_{f} \end{cases}$$





sequences without f



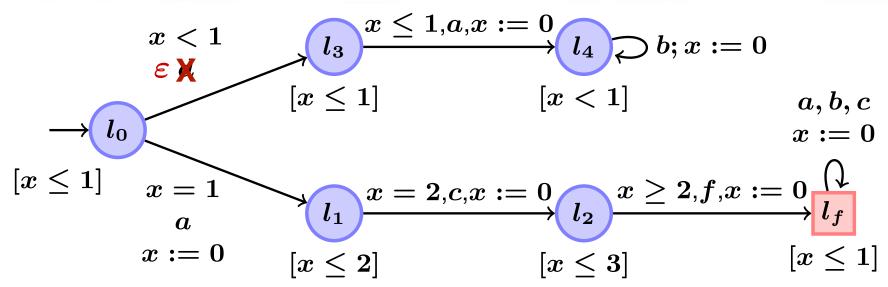


sequences without f

sequences with f

$$1~a~2~c~\delta~f~\cdots$$
 and  $2\leq\delta\leq3$ 





sequences without f

sequences with f

$$1~a~2~c~\delta~f~\cdots$$
 and  $2\leq\delta\leq3$ 

3-predictable

## **Checking Predictability for TA**



Partial observation: projection  $\pi(w)$ 

k-predictor: mapping *P* that satisfies:

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#### **Checking Predictability for TA**



Partial observation: projection  $\pi(w)$ 

Timed languages

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-predictability  $\iff \pi(prefix(L^{\omega}_{\lnot f})) \ \cap \boxed{\pi(L^{-k}_f)} = \varnothing$ 

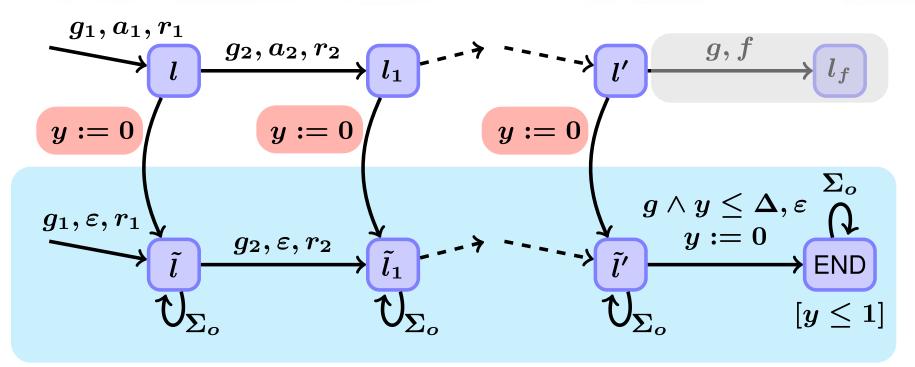
predictability  $\iff$  0-predictability



$$\begin{array}{c|c}
g_1, a_1, r_1 \\
\hline
l \\
\end{matrix} \qquad g_2, a_2, r_2 \\
\hline
l_1 \\
\end{matrix} \qquad \begin{array}{c|c}
g, f \\
\hline
\end{matrix} \qquad \begin{array}{c|c}
\end{matrix} \qquad \qquad \downarrow l_f$$

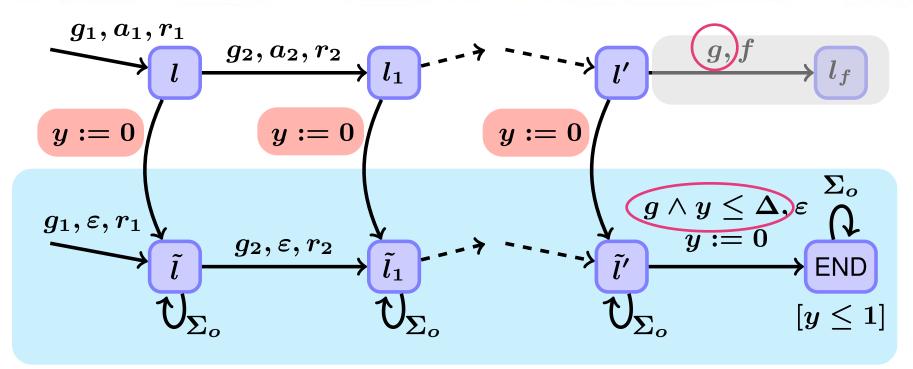






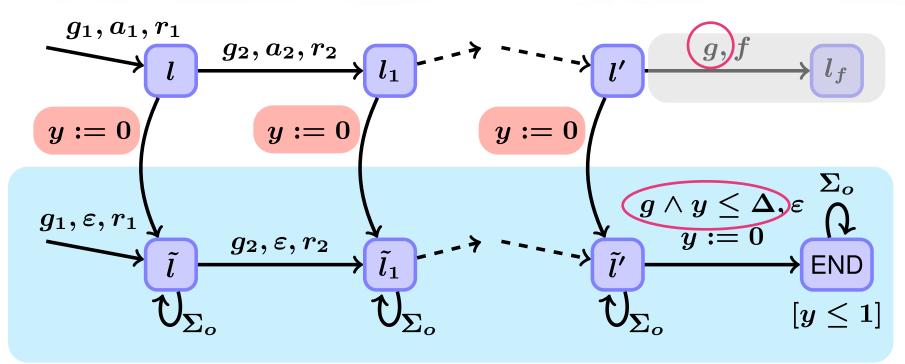
 $\Sigma_o$  = set of observable events





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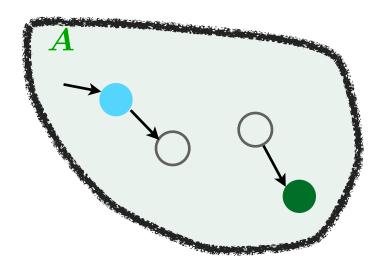
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Büchi emptiness for TA: PSPACE-easy

#### **PSPACE-Hardness**

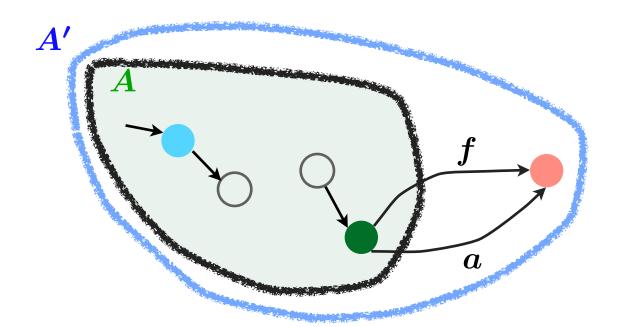




Reachability is PSPACE-hard for TA

#### **PSPACE-Hardness**

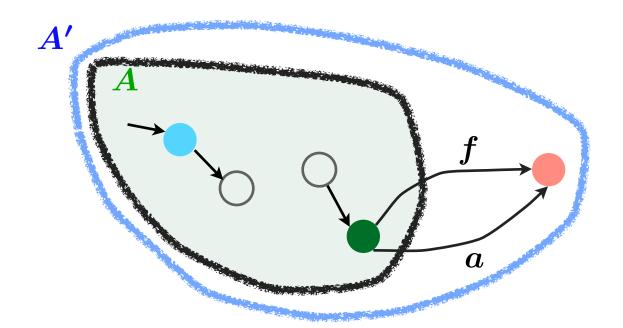




Reachability is PSPACE-hard for TA

### **PSPACE-Hardness**





Reachability is PSPACE-hard for TA

lacksquare is reachable in A iff A' is predictable.

Predictability is PSPACE-hard



#### Predictor should:

- ullet accept the timed language  $\pi(L_f^{-k})$
- be deterministic

Timed Automata are not (always) determinisable

Predictor: online computation of state estimates

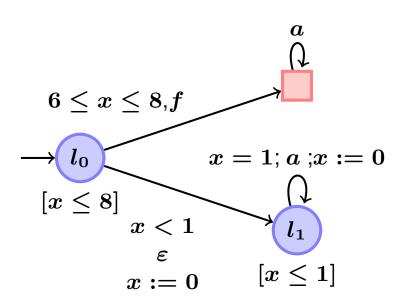


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Predictor: online computation of state estimates



Predictor updates its estimate

- on each observable event
- after a time out

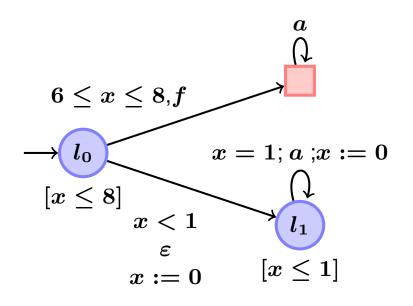


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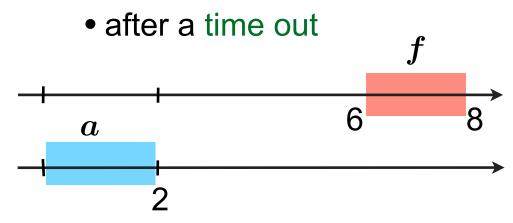
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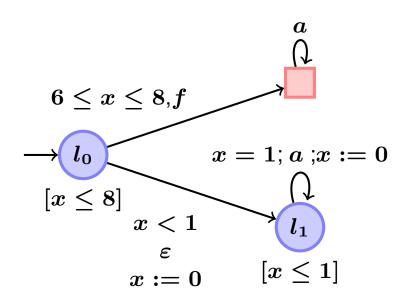


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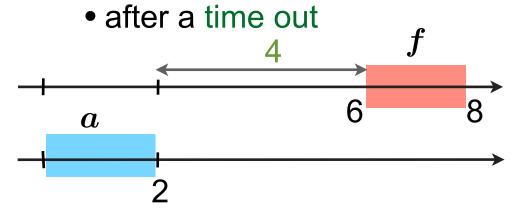
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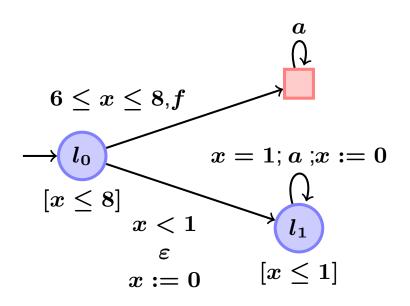


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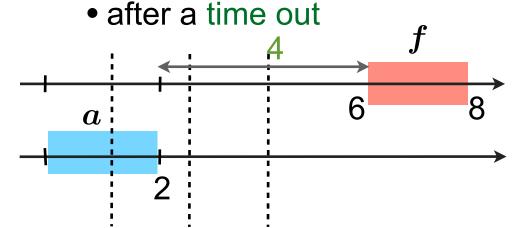
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#### **Sampling Predictability**



Sampling rate  $\alpha \in \mathbb{Q}$ 

 $(k, \alpha)$  – predictor: mapping P that satisfies:

$$\begin{cases} \forall w \in prefix(L^{\omega}_{\neg f} \bmod \alpha), P(\pi(w)) = 0 \\ \forall w \in (L^{-k}_f \bmod \alpha), P(\pi(w)) = 1 \end{cases}$$

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Checking sampling predictability:

impose y := 0 at multiple of  $\alpha$ 

Sampling predictability: there exists a sampling predictor

#### Conclusion



- Simple definition of predictability
- Complexity of predictability for TA
  - PSPACE-complete
- Implementability
  - sampling predictability

- Uniform definition for predictability and diagnosability
- Dynamic observers