### UNIVERSITY OF TWENTE.

Formal Methods & Tools.



# **Confluence Reduction for Markov Automata**



Mark Timmer March 23, 2013



Joint work with Jaco van de Pol and Mariëlle Stoelinga

# The overall goal: efficient and expressive modelling

### Specifying systems with

- ProbabilityDTMCs
- Stochastic timing ← CTMCs

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### Specifying systems with

- Nondeterminism Probabilistic Automata (PAs)
- Probability
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#### Specifying systems with

- Nondeterminism ←
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Interactive Markov Chains (IMCs)

# The overall goal: efficient and expressive modelling

### Specifying systems with

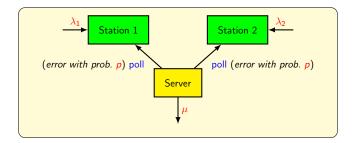
- Nondeterminism
- ProbabilityMarkov Automata (MAs)
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### Specifying systems with

- Nondeterminism 

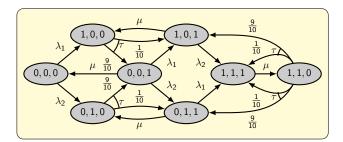
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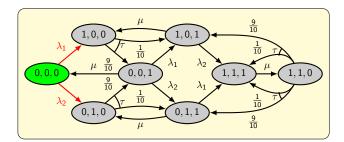
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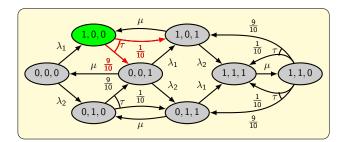
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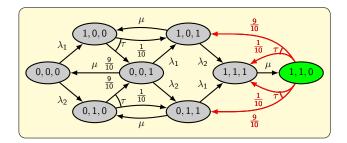
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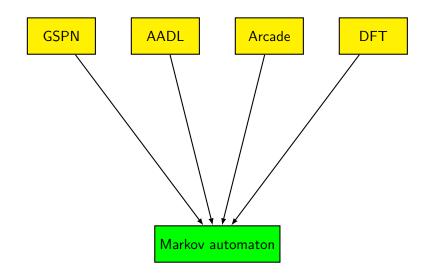
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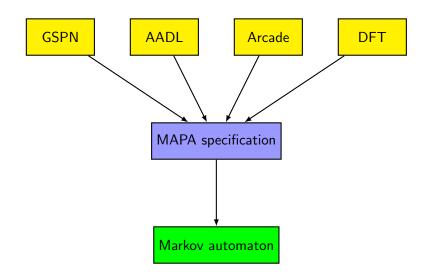
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# Higher-level formalisms that can be mapped to MAs



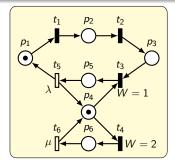
### Higher-level formalisms that can be mapped to MAs



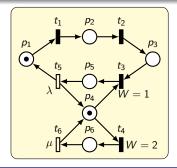
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# Higher-level formalisms mapped to MAs

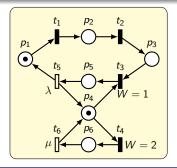


### Higher-level formalisms mapped to MAs



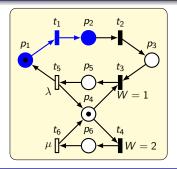
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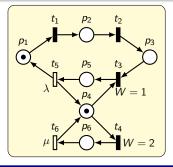


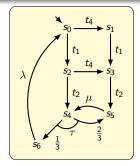
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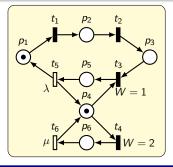


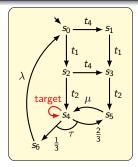
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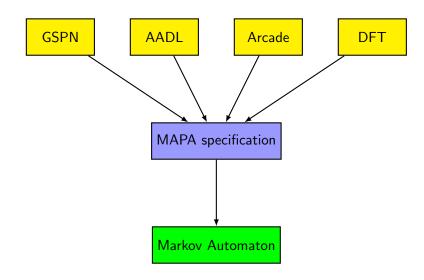


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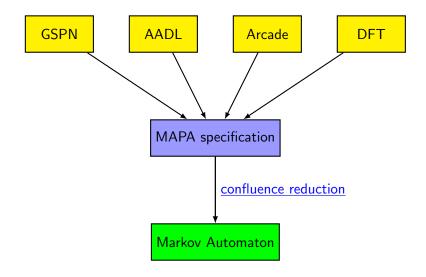


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Introduction

Conclusions



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- Introduction
- 2 Confluence for Markov Automata
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- 4 Symbolic Detection on MAPA Specifications
- 5 Implementation and Case Studies
- 6 Conclusions and Future Work

Case studies

# Invisible transitions connecting equivalent states

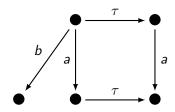
### Invisible transitions in confluence reduction:

- ullet Labelled by au
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Invisible transitions in confluence reduction:

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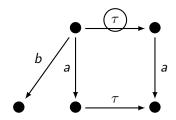
Deterministic  $\tau$ -steps might disable behaviour...



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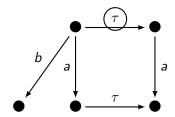
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Deterministic  $\tau$ -steps might disable behaviour... ... though often, they connect equivalent states

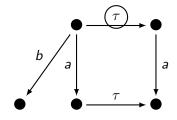


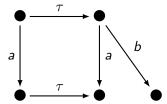
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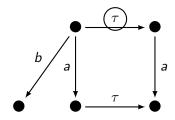


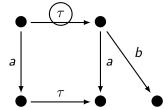
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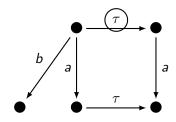


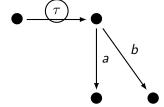
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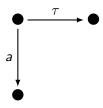
### Non-probabilistic and probabilistic confluence reduction

#### Confluence reduction:

denoting a subset of the invisible transitions as confluent.

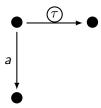
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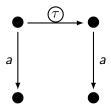
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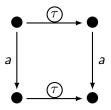
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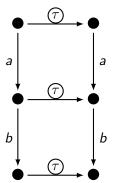
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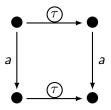
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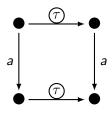
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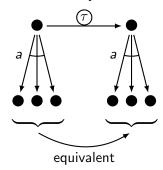
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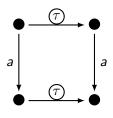


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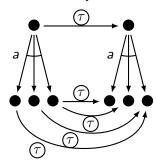
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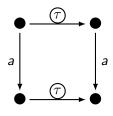


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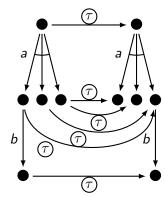
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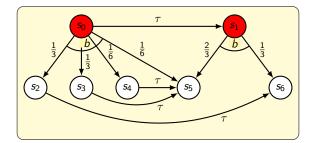
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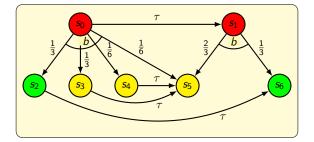
Probabilistically:



### Probabilistic example

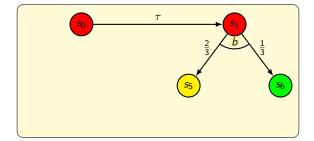


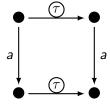
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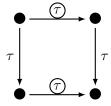


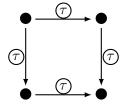
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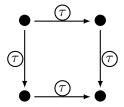
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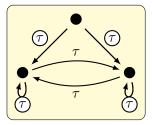


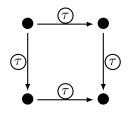


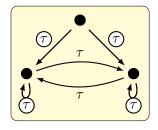


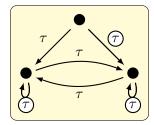


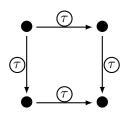


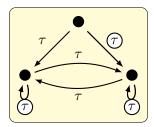


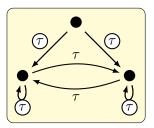


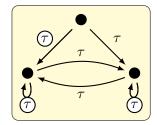




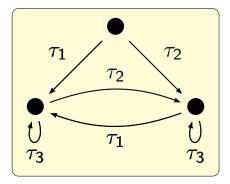






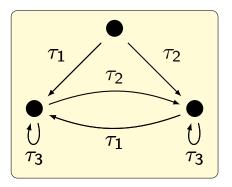


#### Our solution: confluence classification



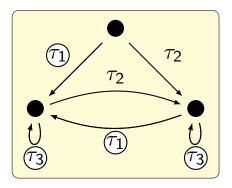
Symbolic detection

#### Our solution: confluence classification



- Mimicking always by a transition from the same group
- For each group, either all transitions or no transitions are confluent

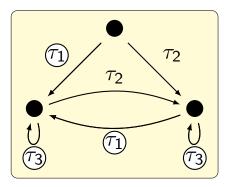
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Symbolic detection

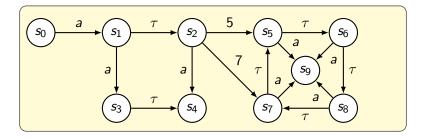
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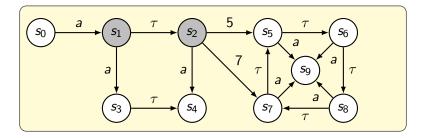


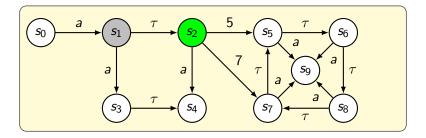
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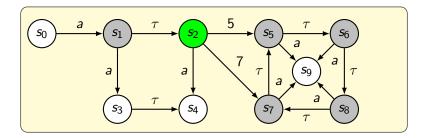
Closure under unions is now really ensured.

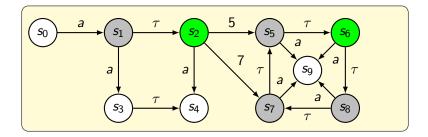


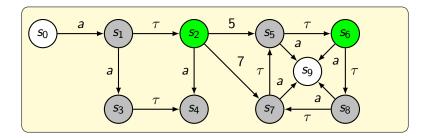
Conclusions

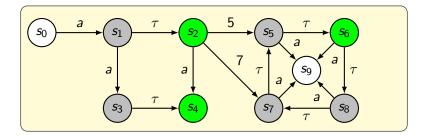






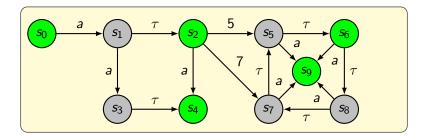


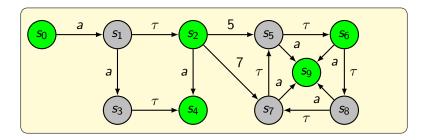


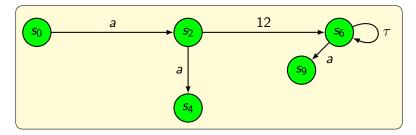


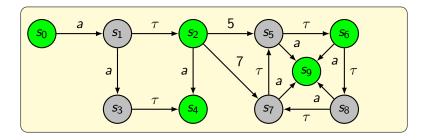
Case studies

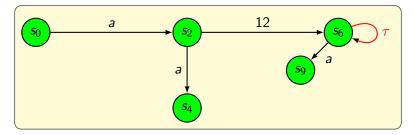
Conclusions











### A process algebra for Markov automata: MAPA

#### Specification language MAPA:

- Based on  $\mu$ CRL (so data), with additional probabilistic choice and Markovian rates
- Semantics defined in terms of Markov automata
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### A process algebra for Markov automata: MAPA

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 Composibility via parallel composition, encapsulation, hiding and renaming

#### **MLPPEs**

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## Example of an MLPPE

#### GSPN-generated MAPA specification

$$\begin{split} \textit{System}(P_1: \mathbb{N}, P_2: \mathbb{N}, P_3: \mathbb{N}, P_4: \mathbb{N}, P_5: \mathbb{N}, P_6: \mathbb{N}) = \\ P_1 \geq 1 \implies \tau \cdot \mathsf{System}(P_1 - 1, P_2 + 1, P_3, P_4, P_5, P_6) \\ + P_2 \geq 1 \implies \tau \cdot \mathsf{System}(P_1, P_2 - 1, P_3 + 1, P_4, P_5, P_6) \\ + P_5 \geq 1 \implies \lambda \cdot \mathsf{System}(P_1 + 1, P_2, P_3, P_4 + 1, P_5 - 1, P_6) \\ + P_6 \geq 1 \implies \mu \cdot \mathsf{System}(P_1, P_2, P_3, P_4 + 1, P_5, P_6 - 1) \\ + (P_3 \geq 1 \land P_4 \geq 1) \lor (P_4 \geq 1) \implies \tau \sum_{i: \{4,5\}} f: \\ \mathsf{System}(P_1, P_2, \mathsf{if}\ i = 4\ \mathsf{then}\ P_3 - 1\ \mathsf{else}\ P_3, P_4 - 1, \\ \mathsf{if}\ i = 4\ \mathsf{then}\ P_5 + 1\ \mathsf{else}\ P_5, \\ \mathsf{if}\ i = 4\ \mathsf{then}\ P_6\ \mathsf{else}\ P_6 + 1) \end{split}$$

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#### $\mathsf{Theorem}$

Every specification (without unguarded recursion) can be linearised to an MLPPE, preserving strong bisimulation.

Confluence State space reduction Symbolic detection

# Detecting confluence symbolically on MLPPEs

Symbolic detection of confluence: denote entire summands to be confluent (i.e., all their concrete transitions are confluent)



Case studies

Conclusions

Symbolic detection of confluence: denote entire summands to be confluent (i.e., all their concrete transitions are confluent)

- Each summand is a group in the confluence classification
- Underapproximation of the actual confluent transitions

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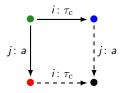
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- It should commute with all the other interactive summands
  - They does not disable each other
  - They should not influence each other's action
  - They should not influence each other's probability expression
  - Their order should not influence the next state

$$X(g:G) = \sum_{\substack{d_i:D_i \\ \cdots}} c_i \Rightarrow \tau \cdot X(n_i)$$

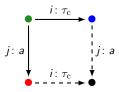
$$\vdots$$

$$+ \sum_{\substack{d_j:D_j \\ c_j \Rightarrow a_j \sum_{e_j:E_j}} f_j \colon X(n_j)$$



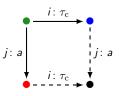
18 / 21

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Heuristics for verifying commutativity of summands i, j:

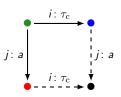
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Heuristics for verifying commutativity of summands i, j:

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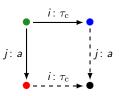
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i: 
$$pc = 3 \Rightarrow \tau \quad \cdot X(pc := 4)$$

$$j: pc = 5 \Rightarrow send(y) \cdot X(pc := 1)$$

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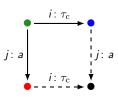
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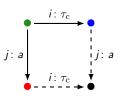
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$$pc1 = 2 \land x > 5 \land y > 2 \Rightarrow \tau \quad X(pc1 := 3, x := 0)$$
  
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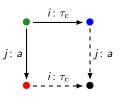
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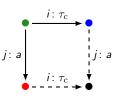
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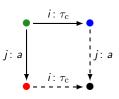
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### We implemented:

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  - Transform GSPNs to MAPA specifications
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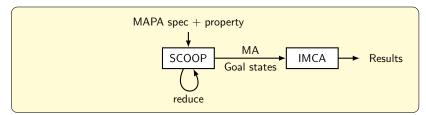
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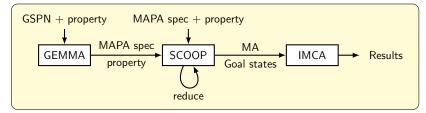
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	Original state space			Reduced state space		Reduction	
Specification	States	Trans.	· IMCA	States Trans	. iMCA	States	Time
leader-3-7	25,505	34,257	103.8	4,652 5,235	5.2	82%	90%
leader-3-9	52,465	71,034	214.3	9,058 10,149	9.9	83%	92%
leader-3-11	93,801	127,683	431.7	15,624 17,463	16.7	83%	93%
leader-4-2	8,467	11,600	74.9	2,071 2,650	5.2	76%	90%
leader-4-3	35,468	50,612	369.3	7,014 8,874	22.4	80%	92%
leader-4-4	101,261	148,024	1,325.3	17,885 22,724	62.2	82%	94%
pol1-2-2-4	4,811	8,578	3.7	3,047 6,814	2.3	37%	32%
pol1-2-2-6	27,651	51,098	90.9	16,557 40,004	49.1	40%	47%
pol1-2-4-2	6,667	11,290	39.9	4,745 9,368	3 26.2	29%	32%
pol1-2-5-2	27,659	47,130	1,573.8	19,721 39,192	2 1,053.5	29%	33%
poll-3-2-2	2,600	4,909	7.1	1,914 4,223	4.8	26%	29%
poll-4-6-1	15,439	29,506	330.0	4,802 18,869	109.3	69%	66%
poll-5-4-1	21,880	43,760	815.0	6,250 28,130	317.5	71%	61%
grid-2	2,508	4,608	2.8	1,393 2,922	2 1.1	44%	49%
grid-3	10,852	20,872	66.3	6,011 13,240	19.8	45%	67%
grid-4	31,832	62,356	922.5	17,565 39,558	316.5	45%	65%

## Conclusions and future work

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- We introduced the first reduction technique for MAs abstracting from internal behaviour: confluence reduction
- It preserves divergences and is closed under unions
- We showed how to detect confluence on MAPA specifications and use the representation map approach to reduce on-the-fly
- Case studies show that significant reductions can be obtained

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#### Future work

- Develop even more powerful reduction techniques
- Define partial-order reduction as a restriction of confluence