

# Compositional Modeling and Minimization of Time-Inhomogeneous Markov Chains

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# Outline

## 1 Motivation

- Continuous-Time Markov Chains
- Inhomogeneous Continuous-Time Markov Chains

## 2 Transient probability distribution

- General case
- Special case

## 3 Inhomogeneous Interactive Markov Chains

## 4 Strong bisimulation

- ICTMC Strong bisimulation
- I<sup>2</sup>MC Strong bisimulation
- Quotienting Algorithm

## 5 Conclusions and Future work

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- Current techniques for evaluation of performance and dependability of computer and communications systems assume time-independence.

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## Time-dependence

- The failure of hardware components is **time-dependent**. Failure rates follow a bath-tub curve.
- Reliability of software in embedded systems is **time-dependent**.
- The process of battery depletion is **time-dependent**.

# CTMC - (Student canteen)

A counter processing requests:

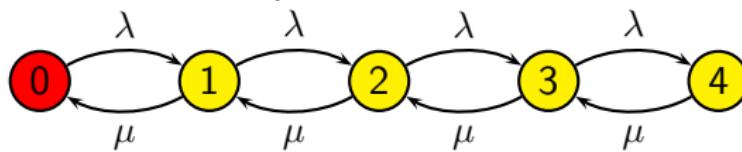


## Configuration

- Each request arrives to the counter with rate  $\lambda$ .
- The counter processes the requests with rate  $\mu$ .
- Additional requests are placed in the **queue**.

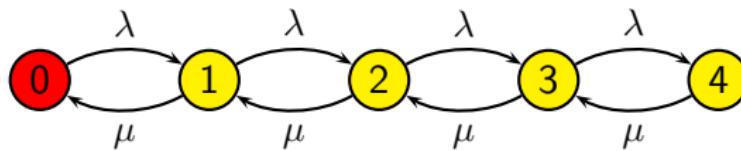
## CTMC - 2 (State of the art)

CTMC model of canteen example:



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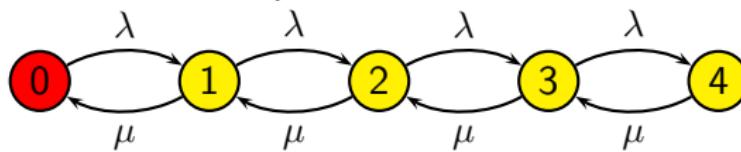


## CTMC results

- **Transient distribution** for CTMCs is well defined.
- Interactive Markov Chain=CTMC+LTS (Labeled Transition Systems).
- Compositional specification of IMCs using **Process Algebra**.
- **Bisimulation** technique for state-space minimization.
- A CTL-like logics for property specification on CTMCs.

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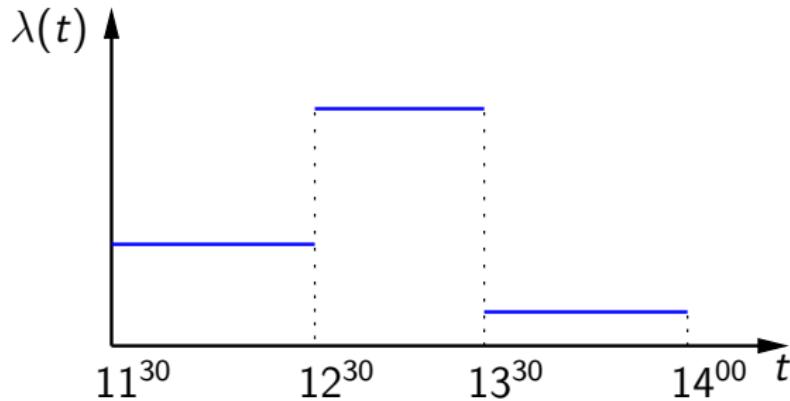
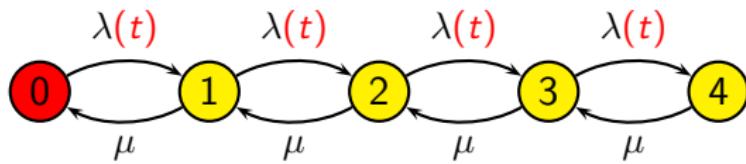


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? - is a model with **time-varying** rates.

## CTMC - 3 (State of the art)



## Definition

An **Inhomogeneous Continuous-Time Markov Chain** (ICTMC) is a tuple  $\mathcal{C} = (\mathbb{S}, \mathbf{R})$  where:

- $\mathbb{S} = \{1, 2, \dots, n\}$  is a countable set of states, and
- $\mathbf{R}(t) = [R_{i,j}(t) \geq 0] \in \mathbb{R}_+^{n \times n}$  is a time-dependent rate matrix, where  $R_{i,j}(t)$  is the rate between states  $i, j \in \mathbb{S}$  at time  $t \geq 0$ .

- $\mathbf{E}(t) = \text{diag}[E_i(t)] \in \mathbb{R}_+^{n \times n}$  is the **exit rate** diagonal matrix, with  $E_i(t) = \sum_{j \in \mathbb{S}} R_{i,j}(t)$   $i, j \in \mathbb{S}$  and  $i \neq j$ .

## ICTMC - 2 (Measures)

- Probability to **leave** some state  $s$  in  $\Delta t$  units of time at time  $t$ :

$$\underbrace{1 - e^{- \int_0^{\Delta t} E_s(t+\ell) d\ell}}_{\text{ICTMC}}$$

$$\underbrace{1 - e^{- E_s \Delta t}}_{\text{CTMC}}$$

- Probability to **select** transition  $s \rightarrow s'$  at time  $t$ :

$$\underbrace{\int_0^{\infty} R_{s,s'}(t+\tau) e^{- \int_0^{\tau} E_s(t+\ell) d\ell} d\tau}_{\text{ICTMC}}$$

$$\underbrace{\frac{R_{s,s'}}{E_s}}_{\text{CTMC}}$$

- Probability to **make** transition  $s \rightarrow s'$  in  $\Delta t$  units of time at time  $t$ :

$$\underbrace{\int_0^{\Delta t} R_{s,s'}(t+\tau) e^{- \int_0^{\tau} E_s(t+\ell) d\ell} d\tau}_{\text{ICTMC}}$$

$$\underbrace{\frac{R_{s,s'}}{E_s} (1 - e^{- E_s \Delta t})}_{\text{CTMC}}$$

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# Transient distribution - (General case)

## Definition

Transient probability distribution -  $\Pr\{X(t + \Delta t) = j\}$  denoted by  $\pi_j(t + \Delta t)$  is the probability to be in a state  $j$  at time  $t + \Delta t$ :

$$\pi_j(t + \Delta t) = \sum_{i \in \mathbb{S}} \Pr\{X(t) = i\} \cdot \Pr\{X(t + \Delta t) = j | X(t) = i\}$$

Transient probability distribution in **matrix form**:

$$\pi(t + \Delta t) = \pi(t) \Phi(t + \Delta t, t),$$

- $\pi(t) = [\pi_1(t), \dots, \pi_n(t)]$  and
- $\Phi(t + \Delta t, t)$  - **transition probability matrix**.

## Transient distribution - 2 (General case)

Transient probability distribution as system of **ODEs**:

$$\frac{d\pi(t)}{dt} = \lim_{\Delta t \rightarrow 0} \frac{\pi(t + \Delta t) - \pi(t)}{\Delta t} = \pi(t) \underbrace{\lim_{\Delta t \rightarrow 0} \frac{[\Phi(t + \Delta t, t) - \mathbf{I}]}{\Delta t}}_{\mathbf{Q}(t)}.$$

- **Infinitesimal generator  $\mathbf{Q}(t)$ :**

$$\mathbf{Q}(t) = \mathbf{R}(t) - \mathbf{E}(t).$$

## Transient distribution - 3 (General case)

- The **solution**  $\pi(t)$ :

$$\pi(t) = \pi(t_0) \Phi(t, t_0)$$

- The general form of  $\Phi(t, t_0)$  is given by the **Peano-Baker** series:

$$\Phi(t, t_0) = \mathbf{I} + \int_{t_0}^t \mathbf{Q}(\tau_1) d\tau_1 + \int_{t_0}^t \mathbf{Q}(\tau_1) \int_{t_0}^{\tau_1} \mathbf{Q}(\tau_2) d\tau_2 d\tau_1 + \dots$$

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- Consider the case of matrix commutativity:

$$\int_{t_0}^t \mathbf{Q}(\tau_1) \int_{t_0}^{\tau_1} \mathbf{Q}(\tau_2) d\tau_2 d\tau_1 = \int_{t_0}^t \int_{t_0}^{\tau_1} \mathbf{Q}(\tau_2) d\tau_2 \mathbf{Q}(\tau_1) d\tau_1$$

- $\Phi(t, t_0)$  takes the form:

$$\Phi(t, t_0) = e^{\int_{t_0}^t \mathbf{Q}(\tau) d\tau}$$

## Transient distribution - 4 (Special Case)

Piecewise uniform rate matrix  $\mathbf{R}(t)$ .

- $M + 1$  - total number of pieces.
- For all  $t \in [t_k, t_{k+1})$  and  $k \leq M \in \mathbb{N}$ :

$$\mathbf{R}(t) = \mathbf{R}_k(t) = f_k(t) \mathbf{R}_k,$$
$$\mathbf{Q}(t) = \mathbf{Q}_k(t) = f_k(t) \mathbf{Q}_k.$$

- Transient probability distribution:

$$\pi(t) = \begin{cases} \pi(t_0) e^{\mathbf{Q}_0 \int_{t_0}^t f_0(\tau) d\tau} & , \text{if } t \in [t_0, t_1) \\ \vdots & \vdots \\ \pi(t_M) e^{\mathbf{Q}_M \int_{t_M}^t f_M(\tau) d\tau} & , \text{if } t \in [t_M, \infty) \end{cases}$$

- $\pi(t_k) = \pi(t_{k-1}) e^{\mathbf{Q}_{k-1} \int_{t_{k-1}}^{t_k} f_{k-1}(\tau) d\tau}$ .

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## Compositionality

- Modeling large stochastic systems is difficult.
- The solution is to construct models of simpler components.
- ICTMC+LTS used for **compositional modeling**.
- ICTMC+LTS=I<sup>2</sup>MC.

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- Modeling large stochastic systems is difficult.
- The solution is to construct models of simpler components.
- ICTMC+LTS used for **compositional modeling**.
- ICTMC+LTS=I<sup>2</sup>MC.

## Definition

An **Inhomogeneous Interactive Markov Chain** (I<sup>2</sup>MC) is a collection  $\mathcal{I} = (\mathbb{S}, \text{Act}, \rightarrow, \mathbf{R}, s^0)$  where  $\mathbb{S}$  and  $\mathbf{R}$  are as before,

- **Act** is a set of actions,
- $\rightarrow \subseteq \mathbb{S} \times \text{Act} \times \mathbb{S}$  is a transition relation and
- $s^0 \in \mathbb{S}$  is the initial state.

# I<sup>2</sup>MC - 2 (Process Algebra)

## Grammar

$$P ::= 0 \mid a.P \mid \lambda(t).P \mid P + P \mid P \parallel_A P \mid P \setminus A \mid X := P$$

Operators:

- Sequential composition -  $a.P \xrightarrow{a} P$
- Sequential composition -  $\lambda(t).P \xrightarrow{\lambda(t)} P$
- Choice -  $P + P$ .
- Parallel Composition -  $P \parallel_A P$  (A is the synchronization set).
- Abstraction -  $P \setminus A$  (A is the abstraction set).
- Recursion -  $X := E[X]$  (E is an expression).

# I<sup>2</sup>MC - 3 (SOS rules)

Structural operational semantic (SOS) rules:

$$\frac{}{a.P \xrightarrow{a} P}$$

$$\frac{P \xrightarrow{a} P'}{P + Q \xrightarrow{a} P'}$$

$$\frac{P \xrightarrow{a} P'}{P \|_A Q \xrightarrow{a} P' \|_A Q} (a \notin A)$$

$$\frac{P \xrightarrow{a} P' \quad \text{and} \quad Q \xrightarrow{a} Q'}{P \|_A Q \xrightarrow{a} P' \|_A Q'} (a \in A)$$

$$\frac{P \xrightarrow{a} P'}{P \setminus A \xrightarrow{a} P \setminus A} (a \notin A)$$

$$\frac{P \xrightarrow{a} P'}{P \setminus A \xrightarrow{\tau} P \setminus A} (a \in A)$$

$$\frac{}{\lambda(t).P \xrightarrow{\lambda(t)} P}$$

$$\frac{P \xrightarrow{\lambda(t)} P'}{P + Q \xrightarrow{\lambda(t)} P'}$$

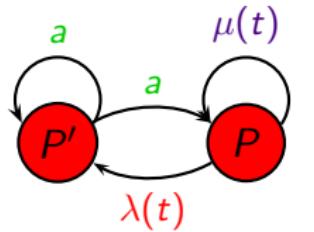
$$\frac{P \xrightarrow{\lambda(t)} P'}{P \|_A Q \xrightarrow{\lambda(t)} P' \|_A Q}$$

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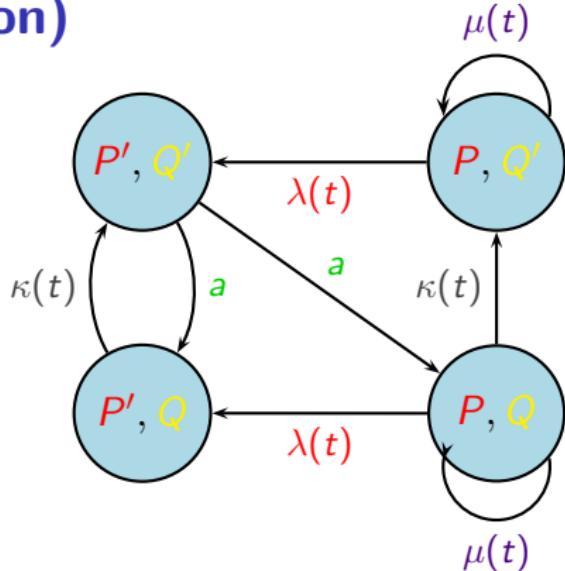
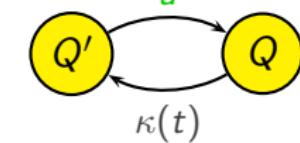
$$\frac{E[X := E/X] \xrightarrow{\lambda(t)} E'}{X := E \xrightarrow{\lambda(t)} E'}$$

$$\frac{E[X := E/X] \xrightarrow{a} E'}{X := E \xrightarrow{a} E'}$$

# I<sup>2</sup>MC - 4 (Parallel composition)



$\parallel$



$$\frac{P \xrightarrow{\lambda(t)} P'}{P \parallel_A Q \xrightarrow{\lambda(t)} P' \parallel_A Q}$$

$$\frac{Q \xrightarrow{\kappa(t)} Q'}{P \parallel_A Q \xrightarrow{\kappa(t)} P \parallel_A Q'}$$

## Memoryless property

$$\Pr\{W(\textcolor{teal}{t}) \leq \textcolor{red}{t}' + \Delta t | W(\textcolor{teal}{t}) > \textcolor{red}{t}'\} = \Pr\{W(\textcolor{teal}{t} + \textcolor{red}{t}') \leq \Delta t\}$$

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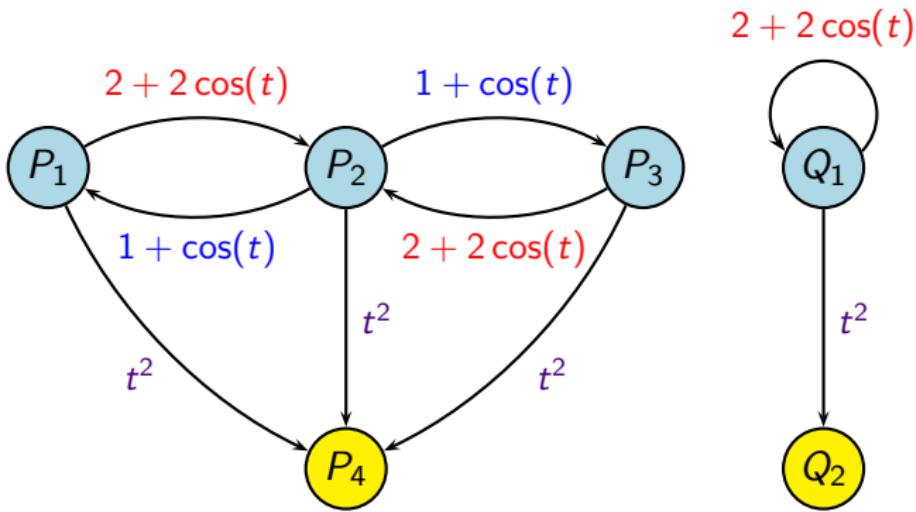
An equivalence relation  $\mathcal{R} \subseteq \mathbb{S} \times \mathbb{S}$  is a **strong bisimulation** whenever: for all  $(P, Q) \in \mathcal{R}$ ,  $t \in \mathbb{R}_{\geq 0}$  and  $C \in \mathbb{S}/\mathcal{R}$ :

$$R(P, C, t) = R(Q, C, t),$$

where

- $R$  is the multiset  $R(P, C, t) = \sum_i \{|\lambda(t)|P \xrightarrow{i} P', P' \in C|\}$ .
- $P$  and  $Q$  are strongly bisimilar, denoted  $P \sim Q$ , if they are contained in some strong bisimulation  $\mathcal{R}$ , i.e.  $(P, Q) \in \mathcal{R}$ .

## ICTMC Strong bisimulation - 2 (Example)



Equivalence classes:

- $C_1 = \{P_1, P_2, P_3, Q_1\}$
- $C_2 = \{P_4, Q_2\}$

## ICTMC Strong bisimulation - 3 (Quotient)

### Definition

For the ICTMC  $\mathcal{C} = (\mathbb{S}, \mathbf{R}, s^0)$  and  $\sim$ , the quotient  $\mathcal{C}/\sim$  is defined by  $\mathcal{C}/\sim = (\mathbb{S}/\sim, \mathbf{R}_\sim, s_\sim^0)$  where  $s_\sim^0 = [s^0]_\sim$  and  $\mathbf{R}_\sim$  is defined by:

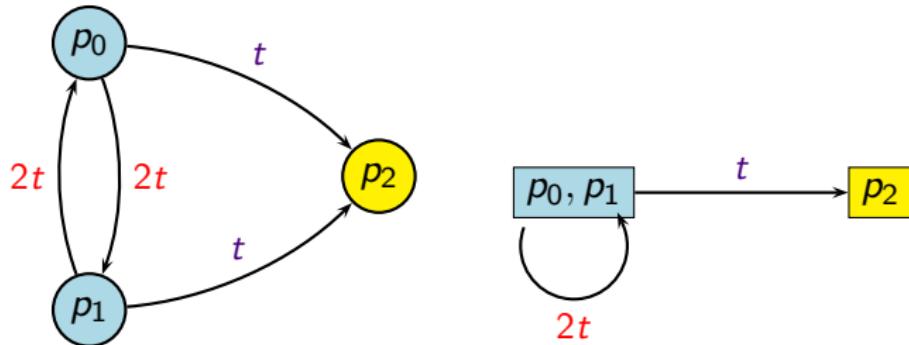
$$R([P]_\sim, [P']_\sim, t) = \frac{\sum_{P'' \in [P]_\sim} R(P'', [P']_\sim, t)}{|[P]_\sim|} \text{ for all } t \in \mathbb{R}_{\geq 0},$$

and  $|[P]_\sim|$  the size of  $[P]_\sim$ .

$$\pi_C(t) = \sum_{s \in C} \pi_s(t) \quad \text{for all } t \in \mathbb{R}_{\geq 0},$$

- $\pi_C(t)$  - transient distribution of state  $C$  in quotient  $\mathcal{C}/\sim$ .

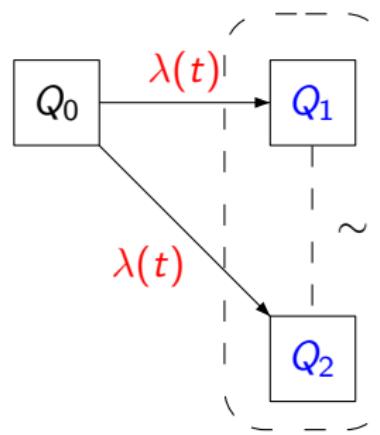
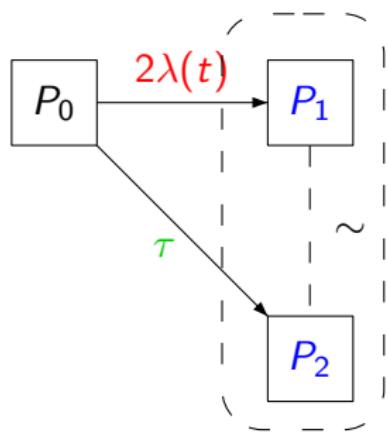
## ICTMC Strong bisimulation - 4 (Example)



Equivalence classes:

- $C_1 = \{p_0, p_1\}$
- $C_2 = \{p_2\}$

# I<sup>2</sup>MC Strong bisimulation



## Maximal progress

- Internal action  $τ$  consumes no time.
- Probability to take  $P_0 \xrightarrow{2\lambda(t)} P_1$  in zero time is 0.

# I<sup>2</sup>MC Strong bisimulation - 2 (Definition)

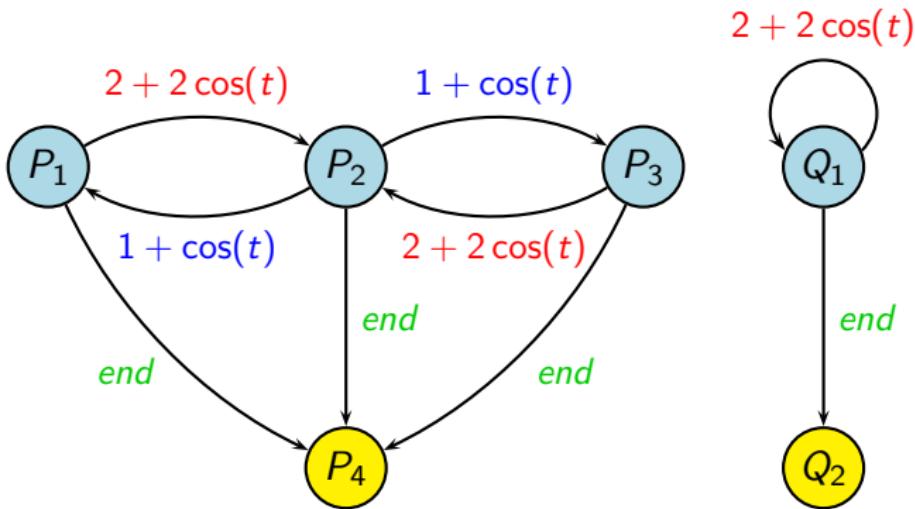
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- $P \xrightarrow{a} P'$  implies  $Q \xrightarrow{a} Q'$  for some  $Q'$  and  $(P', Q') \in \mathcal{R}$ .
- $Q \xrightarrow{a} Q'$  implies  $P \xrightarrow{a} P'$  for some  $P'$  and  $(P', Q') \in \mathcal{R}$ .
- $P \xrightarrow{T} ($ or  $Q \xrightarrow{T}$ ) implies  $R(P, C, t) = R(Q, C, t)$ .

$P$  and  $Q$  are strongly bisimilar, denoted  $P \sim Q$ , if they are contained in some strong bisimulation  $\mathcal{R}$ , i.e.  $(P, Q) \in \mathcal{R}$ .

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Equivalence classes:

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- $C_2 = \{P_4, Q_2\}$

# I<sup>2</sup>MC Strong bisimulation - 4 (Congruence)

Congruence properties:

- $P \sim P'$  implies  $P \parallel_A Q \sim P' \parallel_A Q$ .
- $P \sim Q$  implies  $a.P \sim a.Q$  for any  $a \in \text{Act}$ .
- $P \sim Q$  implies  $\lambda(t).P \sim \lambda(t).Q$  for any  $t \in \mathbb{R}_{\geq 0}$ .
- $P \sim Q$  implies  $P + R \sim Q + R$ .
- $P \sim Q$  implies  $X := P \sim X := Q$ .

# I<sup>2</sup>MC Strong bisimulation - 5 (Axioms)

Sound and complete axioms for sequential fragment:

- $P + 0 = P$
- $(P + Q) + R = P + (Q + R)$
- $a.P + a.P = a.P$  and  $\lambda(t).P + \mu(t).P = (\lambda(t) + \mu(t)).P$
- $\lambda(t).P + \tau.Q = \tau.Q$

# Quotienting Algorithm

- Composition modeling results in **huge** state-spaces.
- Bisimulation minimization achieve **exponential** state-space reduction for LTS and CTMC.
- We adopt the **partition refinement** paradigm for minimizing  $I^2MC$ .

# Quotienting Algorithm

- Composition modeling results in **huge** state-spaces.
- Bisimulation minimization achieve **exponential** state-space reduction for LTS and CTMC.
- We adopt the **partition refinement** paradigm for minimizing  $I^2MC$ .

- For general rate function the problem is **undecidable**.
- We restrict the rate function to **piecewise constant**, **piecewise polynomial** and **polynomial**.

## Quotienting Algorithm - 2

**Input:** I<sup>2</sup>MC  $(\mathbb{S}, \text{Act}, \rightarrow, \mathbf{R})$ ,  $M$

**Output:** I<sup>2</sup>MC  $(\mathbb{S}/\sim, \text{Act}, \rightarrow_{\sim}, \tilde{\mathbf{R}})$

- 1:  $\Pi_{\tau} := \{s \in \mathbb{S} | s \xrightarrow{\tau}\}$     $\Pi := \{s \in \mathbb{S} | s \not\xrightarrow{\tau}\}$
- 2:  $L := \text{push}(\{\mathbb{S}\})$
- 3: **while**  $L \neq \emptyset$  **do**
- 4:      $C := \text{pop}(L)$
- 5:      $[\Pi, L] := \text{Refine}_a(\Pi, C, L, \text{Act})$
- 6:      $[\Pi, L] := \text{Refine}(\Pi, C, L, \mathbf{R}, M)$
- 7: **end while**

Rate matrix parameter  $M$ :

- $M + 1$  - number of constant or polynomial pieces,
- $M + 1$  - degree of the polynomial for polynomial rate function.

# Quotienting Algorithm - 3

**Input:** I<sup>2</sup>MC  $(\mathbb{S}, \text{Act}, \rightarrow, \mathbf{R})$ ,  $M$

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Refinement functions:

- $\text{Refine}_a$  - partition refinement with respect to **actions**,
- $\text{Refine}$  - partition refinement with respect to **rates**.

## Refining wrt. actions

$\text{Refine}_a$  can be implemented using a simple adaptation of Paige and Tarjan's algorithm.

### Complexity

- Time -  $\mathcal{O}(m_a \log n)$ ,
- Space -  $\mathcal{O}(m_a)$ ,

where  $m_a$  is the number of action-labelled transitions.

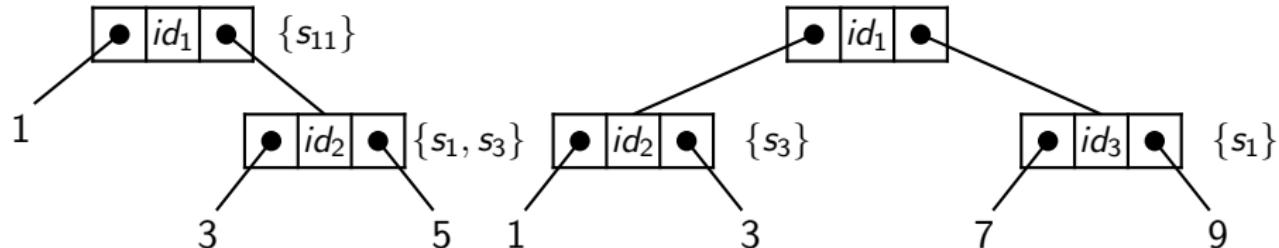
## Refining wrt. rates - 2

**Input:**  $\text{Refine}(\Pi, C, L, R, M)$

```
1: for  $i = 0$  to  $M$  do
2:    $B_\Pi := \emptyset$ 
3:   ... (*initialization*) ...
4:   for all  $s \in \text{Pre}(C)$  do
5:      $B' := [s]_\Pi$ 
6:     delete  $s$  from  $B'$ 
7:     Insert( $\text{T}(B')$ ,  $s$ )
8:     if  $B' \notin B_\Pi$  then
9:        $B_\Pi := B_\Pi \cup B'$ 
10:      end if
11:    end for
12:    for all  $B' \in B_\Pi$  do
13:       $\Pi := \Pi \cup$ 
14:       $\{\text{T}(B')[1], \dots, \text{T}(B')[k]\}$ 
15:    end for
16:  end for
17:
```

- $\text{Pre}(C)$  - predecessors of  $C$ .
- $B_\Pi$  - set of blocks in  $\Pi$ .
- $\text{T}(B')$  - binary tree  $B'$ .
- $\text{T}(B')[j]$  -  $j$ 'th block of  $\text{T}(B')$ .
- *Insert*( $\text{T}(B')$ ,  $s$ ) - insert  $s \in B'$  into  $\text{T}(B')$ .

## Refining wrt. rates - 3



$T(B')$  node structure:

- $node.left$  - points to the left child.
- $node.right$  - points to the right child.
- $node.sum$  - stores the sum of all rates to splitter  $C$  on the  $i$ 'th piece.
- $node.S$  - stores all states with the same  $node.sum$ .

## Refining wrt. rates - 4

Using **splay** trees instead of **balanced** binary trees results in:

- Time complexity -  $\mathcal{O}(Mm_r \log n)$ ,
- Space complexity -  $\mathcal{O}(m_r)$ ,

where  $m_r$  is the number of rate-labelled transitions.

## Refining wrt. rates - 4

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- Space complexity -  $\mathcal{O}(m_r)$ ,

where  $m_r$  is the number of rate-labelled transitions.

### Bisimulation Quotienting Algorithm

- Time complexity -  $\mathcal{O}(m_a \log n + Mm_r \log n)$ ,
- Space complexity -  $\mathcal{O}(m_a + m_r)$ .

## 1 Motivation

- Continuous-Time Markov Chains
- Inhomogeneous Continuous-Time Markov Chains

## 2 Transient probability distribution

- General case
- Special case

## 3 Inhomogeneous Interactive Markov Chains

## 4 Strong bisimulation

- ICTMC Strong bisimulation
- I<sup>2</sup>MC Strong bisimulation
- Quotienting Algorithm

## 5 Conclusions and Future work

# Conclusions and Future work

## Conclusions:

- Full-fledged process algebra for interactive ICTMCs.
- Congruence results for weak and strong bisimulation.
- Polynomial-time quotienting algorithm.

## Future work:

- Extension of quotienting algorithm to a larger class of rate functions.
- Simulation relations for ICTMCs.
- Model-checking algorithms for ICTMCs.

Thank you!