

# Lecture 9

## Continuous-time Markov chains...

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# Overview

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- Transient probabilities
  - uniformisation
- Steady-state probabilities
- CSL: Continuous Stochastic Logic
  - syntax
  - semantics
  - examples

# Recall

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- Continuous-time Markov chain:  $C = (S, s_{\text{init}}, R, L)$ 
  - $R : S \times S \rightarrow \mathbb{R}_{\geq 0}$  is the **transition rate matrix**
  - rates interpreted as parameters of exponential distributions

- Embedded DTMC:  $\text{emb}(C) = (S, s_{\text{init}}, P^{\text{emb}(C)}, L)$

$$P^{\text{emb}(C)}(s, s') = \begin{cases} R(s, s')/E(s) & \text{if } E(s) > 0 \\ 1 & \text{if } E(s) = 0 \text{ and } s = s' \\ 0 & \text{otherwise} \end{cases}$$

- Infinitesimal generator matrix

$$Q(s, s') = \begin{cases} R(s, s') & s \neq s' \\ -\sum_{s \neq s'} R(s, s') & \text{otherwise} \end{cases}$$

# Transient and steady-state behaviour

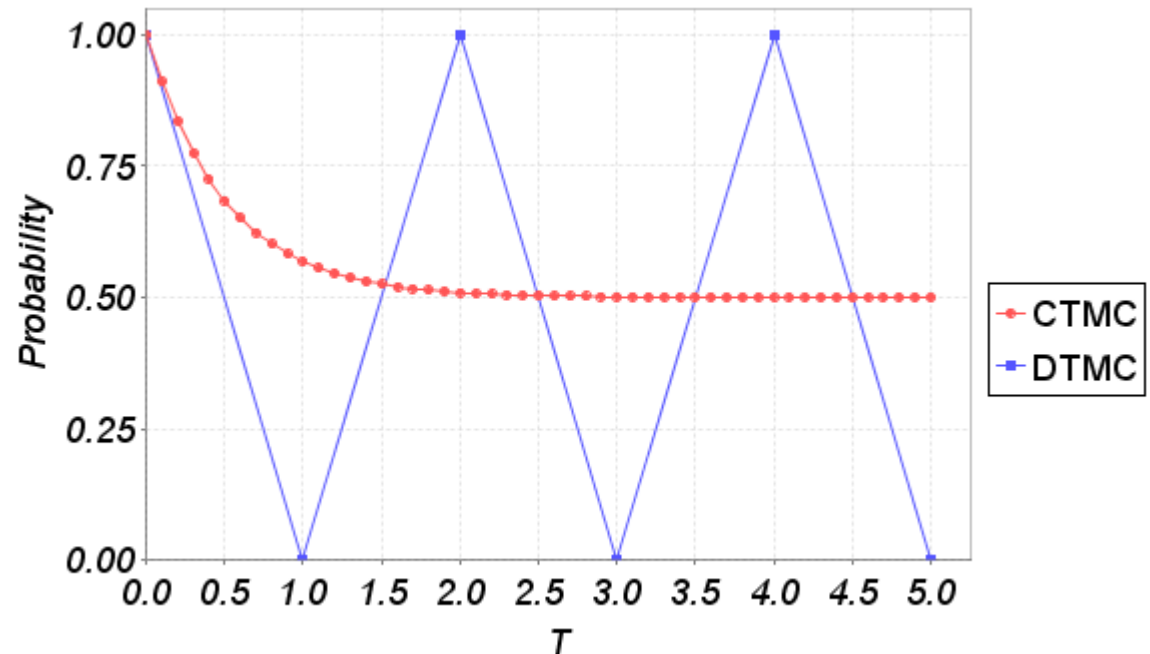
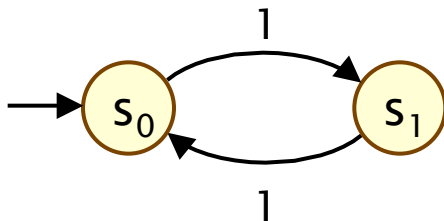
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- Transient behaviour
  - state of the model at a particular **time instant**
  - $\underline{\pi}_{s,t}^C(s')$  is probability of, having started in state  $s$ , being in state  $s'$  at time  $t$  (in CTMC  $C$ )
  - $\underline{\pi}_{s,t}^C(s') = \Pr_s\{ \omega \in \text{Path}^C(s) \mid \omega@t=s' \}$
- Steady-state behaviour
  - state of the model in the **long-run**
  - $\underline{\pi}_s^C(s')$  is probability of, having started in state  $s$ , being in state  $s'$  in the long run
  - $\underline{\pi}_s^C(s') = \lim_{t \rightarrow \infty} \underline{\pi}_{s,t}^C(s')$
  - intuitively: long-run percentage of time spent in each state

# Computing transient probabilities

- Consider a simple example
  - and compare to the case for DTMCs
- What is the probability of being in state  $s_0$  at time  $t$ ?

- DTMC/CTMC:



# Computing transient probabilities

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- $\Pi_t$  – matrix of transient probabilities
  - $\Pi_t(s, s') = \underline{\pi}_{s,t}(s')$
- $\Pi_t$  solution of the differential equation:  $\Pi_t' = \Pi_t \cdot Q$ 
  - where  $Q$  is the infinitesimal generator matrix
- Can be expressed as a **matrix exponential** and therefore evaluated as a **power series**

$$\Pi_t = e^{Q \cdot t} = \sum_{i=0}^{\infty} (Q \cdot t)^i / i!$$

- computation potentially **unstable**
- probabilities instead computed using **uniformisation**

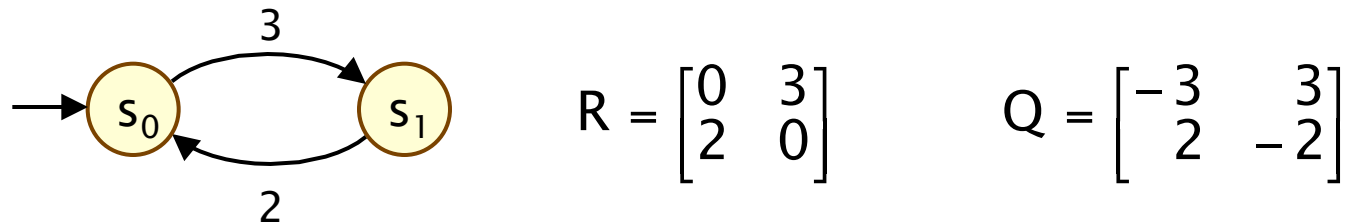
# Uniformisation

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- We build the **uniformised DTMC**  $\text{unif}(C)$  of CTMC  $C$
- If  $C = (S, s_{\text{init}}, R, L)$ , then  $\text{unif}(C) = (S, s_{\text{init}}, P^{\text{unif}(C)}, L)$ 
  - set of states, initial state and labelling the same as  $C$
  - $P^{\text{unif}(C)} = I + Q/q$
  - $I$  is the  $|S| \times |S|$  identity matrix
  - $q \geq \max \{ E(s) \mid s \in S \}$  is the **uniformisation rate**
- Each time step (epoch) of uniformised DTMC corresponds to **one exponentially distributed delay with rate  $q$** 
  - if  $E(s)=q$  transitions the same as embedded DTMC (residence time has the same distribution as one epoch)
  - if  $E(s)<q$  add self loop with probability  $1-E(s)/q$  (residence time longer than  $1/q$  so one epoch may not be ‘long enough’)

# Uniformisation – Example

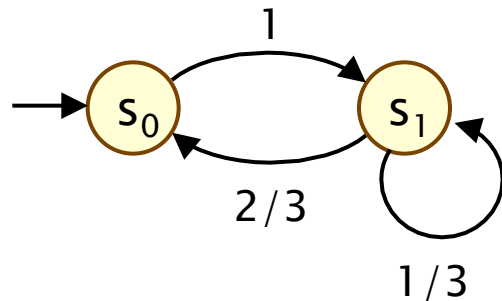
- CTMC C:



- Uniformised DTMC  $\text{unif}(C)$

– let uniformisation rate  $q = \max_s \{ E(s) \} = 3$

$$P^{\text{unif}(C)} = I + Q / q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} -1 & 1 \\ 2/3 & -2/3 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 2/3 & 1/3 \end{bmatrix}$$





# Uniformisation

- Using the uniformised DTMC the transient probabilities can be expressed by:

$$\begin{aligned}\Pi_t &= e^{Q \cdot t} = e^{q \cdot (P^{\text{unif}(C)} - I) \cdot t} = e^{(q \cdot t) \cdot P^{\text{unif}(C)}} \cdot e^{-q \cdot t} \\ &= e^{-q \cdot t} \cdot \left( \sum_{i=0}^{\infty} \frac{(q \cdot t)^i}{i!} \cdot \left( P^{\text{unif}(C)} \right)^i \right) \\ &= \sum_{i=0}^{\infty} \left( e^{-q \cdot t} \cdot \frac{(q \cdot t)^i}{i!} \right) \left( P^{\text{unif}(C)} \right)^i \\ &= \sum_{i=0}^{\infty} \gamma_{q \cdot t, i} \cdot \left( P^{\text{unif}(C)} \right)^i\end{aligned}$$

$\gamma_{q \cdot t, i}$  is Poisson probability with parameter  $q \cdot t$

$P^{\text{unif}(C)}$  is stochastic (all entries in  $[0,1]$  & rows sum to 1); therefore computations with  $P$  are more numerically stable than  $Q$

# Uniformisation

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$$\Pi_t = \sum_{i=0}^{\infty} Y_{q \cdot t, i} \cdot \left( P^{\text{unif}(C)} \right)^i$$

- $(P^{\text{unif}(C)})^i$  is probability of jumping between each pair of states **in  $i$  steps**
- $Y_{q \cdot t, i}$  is the  **$i$ th Poisson probability** with parameter  $q \cdot t$ 
  - the probability of  $i$  steps occurring in time  $t$ , given each has delay exponentially distributed with rate  $q$
- Can **truncate** the (infinite) summation using the techniques of Fox and Glynn [FG88], which allow **efficient computation** of the Poisson probabilities

# Uniformisation

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- Computing  $\underline{\pi}_{s,t}$  for a fixed state  $s$  and time  $t$ 
  - can be computed **efficiently** using **matrix-vector operations**
  - pre-multiply the matrix  $\Pi_t$  by the initial distribution
  - in this case:  $\underline{\pi}_{s,0}(s')$  equals 1 if  $s=s'$  and 0 otherwise

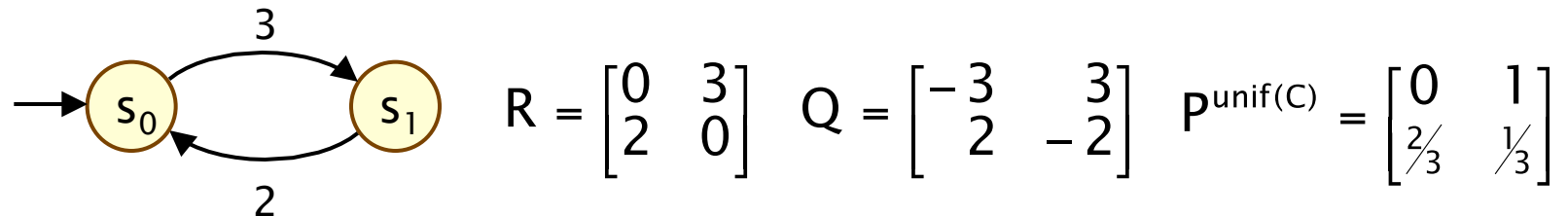
$$\begin{aligned}\underline{\pi}_{s,t} &= \underline{\pi}_{s,0} \cdot \Pi_t = \underline{\pi}_{s,0} \cdot \sum_{i=0}^{\infty} Y_{q \cdot t, i} \cdot \left( P^{\text{unif}(C)} \right)^i \\ &= \sum_{i=0}^{\infty} Y_{q \cdot t, i} \cdot \underline{\pi}_{s,0} \cdot \left( P^{\text{unif}(C)} \right)^i\end{aligned}$$

- compute iteratively to avoid the computation of matrix powers

$$\left( \underline{\pi}_{s,t} \cdot P^{\text{unif}(C)} \right)^{i+1} = \left( \underline{\pi}_{s,t} \cdot P^{\text{unif}(C)} \right)^i \cdot P^{\text{unif}(C)}$$

# Uniformisation – Example

- CTMC C, uniformised DTMC for  $q=3$



- Initial distribution:  $\underline{\pi}_{s_0,0} = [1, 0]$
- Transient probabilities for time  $t = 1$ :

$$\begin{aligned} \underline{\pi}_{s_0,1} &= \sum_{i=0}^{\infty} Y_{q \cdot t, i} \cdot \underline{\pi}_{s_0,0} \cdot \left( P^{\text{unif}(C)} \right)^i \\ &= Y_{3,0} \cdot [1, 0] \cdot \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + Y_{3,1} \cdot [1, 0] \cdot \begin{bmatrix} 0 & 1 \\ 2/3 & 1/3 \end{bmatrix} + Y_{3,2} \cdot [1, 0] \cdot \begin{bmatrix} 0 & 1 \\ 2/3 & 1/3 \end{bmatrix}^2 + \dots \\ &\approx [0.404043, 0.595957] \end{aligned}$$

# Steady-state probabilities

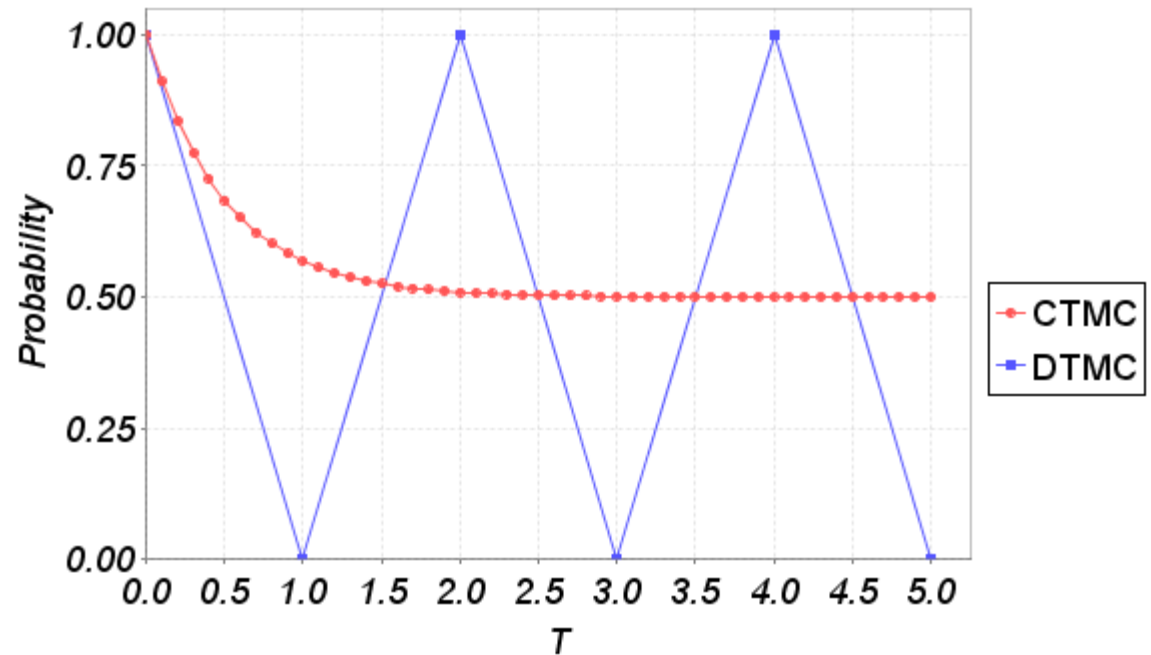
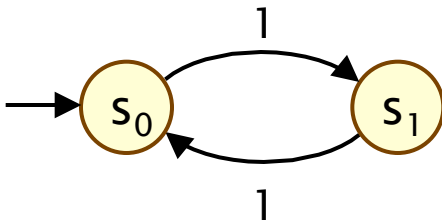
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- Limit  $\underline{\pi}_s^C(s') = \lim_{t \rightarrow \infty} \underline{\pi}_{s,t}^C(s')$ 
  - exists for all finite CTMCs
  - (see next slide)
- As for DTMCs, need to consider the underlying graph structure of the Markov chain:
  - reachability (between pairs) of states
  - bottom strongly connected components (BSCCs)
  - one special case to consider: absorbing states are BSCCs
  - note: can do this equivalently on embedded DTMC
- CTMC is **irreducible** if all its states belong to a single BSCC; otherwise reducible

# Periodicity

- Unlike for DTMCs, do not need to consider periodicity
- e.g. probability of being in state  $s_0$  at time  $t$ ?

- DTMC/CTMC:



# Irreducible CTMCs

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- For an irreducible CTMC:
  - the steady-state probabilities are **independent of the starting state**: denote the steady state probabilities by  $\underline{\pi}^C(s')$

- These probabilities can be computed as
  - the **unique solution of the linear equation system**:

$$\underline{\pi}^C \cdot Q = \underline{0} \quad \text{and} \quad \sum_{s \in S} \underline{\pi}^C(s) = 1$$

where  $Q$  is the infinitesimal generator matrix of  $C$

- Solved by standard means:
  - direct methods, such as Gaussian elimination
  - iterative methods, such as Jacobi and Gauss–Seidel

# Balance equations

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$$\underline{\pi}^C \cdot Q = \underline{0} \quad \text{and} \quad \sum_{s \in S} \underline{\pi}^C(s) = 1$$

balance the rate of  
leaving and entering  
a state

normalisation

For all  $s \in S$ :

$$\underline{\pi}^C(s) \cdot (-\sum_{s' \neq s} R(s, s')) + \sum_{s' \neq s} \underline{\pi}^C(s') \cdot R(s', s) = 0$$

$\Leftrightarrow$

$$\underline{\pi}^C(s) \cdot \sum_{s' \neq s} R(s, s') = \sum_{s' \neq s} \underline{\pi}^C(s') \cdot R(s', s)$$

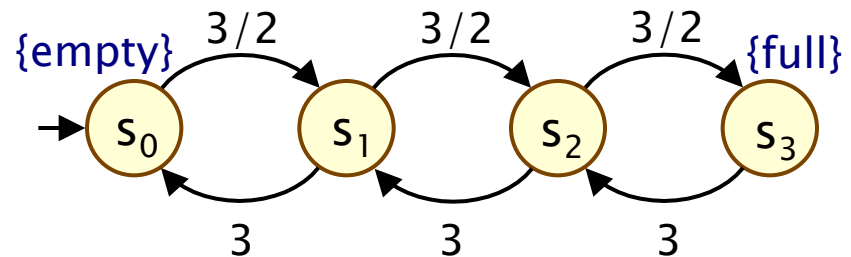
Equivalent to:  $\underline{\pi}^C \cdot P = \underline{\pi}^C$  where  $P$  is matrix for embedded DTMC



# Steady-state – Example

- Solve:  $\underline{\pi} \cdot \mathbf{Q} = 0$  and  $\sum \underline{\pi}(s) = 1$

$$\mathbf{Q} = \begin{bmatrix} -3/2 & 3/2 & 0 & 0 \\ 3 & -9/2 & 3/2 & 0 \\ 0 & 3 & -9/2 & 3/2 \\ 0 & 0 & 3 & -3 \end{bmatrix}$$



$$-3/2 \cdot \underline{\pi}(s_0) + 3 \cdot \underline{\pi}(s_1) = 0$$

$$3/2 \cdot \underline{\pi}(s_0) - 9/2 \cdot \underline{\pi}(s_1) + 3 \cdot \underline{\pi}(s_2) = 0$$

$$3/2 \cdot \underline{\pi}(s_1) - 9/2 \cdot \underline{\pi}(s_2) + 3 \cdot \underline{\pi}(s_3) = 0$$

$$3/2 \cdot \underline{\pi}(s_2) - 3 \cdot \underline{\pi}(s_3) = 0$$

$$\underline{\pi}(s_0) + \underline{\pi}(s_1) + \underline{\pi}(s_2) + \underline{\pi}(s_3) = 1$$

$$\underline{\pi} = [ 8/15, 4/15, 2/15, 1/15 ]$$

# Reducible CTMCs

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- For a reducible CTMC:
  - the steady-state probabilities  $\underline{\pi}^C(s')$  depend on start state  $s$
- Find all BSCCs of CTMC, denoted  $\text{bscc}(C)$
- Compute:
  - steady-state probabilities  $\underline{\pi}^T$  of sub-CTMC for each BSCC  $T$
  - probability  $\text{ProbReach}^{\text{emb}(C)}(s, T)$  of reaching each  $T$  from  $s$
- Then:

$$\underline{\pi}_s^C(s') = \begin{cases} \text{ProbReach}^{\text{emb}(C)}(s, T) \cdot \underline{\pi}^T(s') & \text{if } s' \in T \text{ for some } T \in \text{bscc}(C) \\ 0 & \text{otherwise} \end{cases}$$

# CSL

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- Temporal logic for describing properties of CTMCs
  - CSL = Continuous Stochastic Logic [ASSB00,BHHK03]
  - extension of (non-probabilistic) temporal logic CTL
- Key additions:
  - probabilistic operator  $P$  (like PCTL)
  - steady state operator  $S$
- Example:  $\text{down} \rightarrow P_{>0.75} [ \neg\text{fail} \mathbf{U}^{[1,2.5]} \text{up} ]$ 
  - when a shutdown occurs, the probability of a system recovery being completed between 1 and 2.5 hours without further failure is greater than 0.75
- Example:  $S_{<0.1} [ \text{insufficient\_routers} ]$ 
  - in the long run, the chance that an inadequate number of routers are operational is less than 0.1

# CSL syntax

- CSL syntax:

- $\phi ::= \text{true} \mid a \mid \phi \wedge \phi \mid \neg \phi \mid P_{\sim p} [\psi] \mid S_{\sim p} [\phi]$  (state formulae)

- $\psi ::= X \phi \mid \phi U^I \phi$

“next”

“time bounded until”

in the “long run”  $\phi$  is true with probability  $\sim p$

- where  $a$  is an atomic proposition,  $I$  interval of  $\mathbb{R}_{\geq 0}$  and  $p \in [0, 1]$ ,  $\sim \in \{<, >, \leq, \geq\}$

- A CSL formula is always a state formula

- path formulae only occur inside the  $P$  operator

# CSL semantics for CTMCs

- CSL formulae interpreted over states of a CTMC
  - $s \models \phi$  denotes  $\phi$  is “true in state  $s$ ” or “satisfied in state  $s$ ”
- Semantics of state formulae:
  - for a state  $s$  of the CTMC  $(S, s_{\text{init}}, R, L)$ :

- |                                    |                                                                  |
|------------------------------------|------------------------------------------------------------------|
| – $s \models a$                    | $\Leftrightarrow a \in L(s)$                                     |
| – $s \models \phi_1 \wedge \phi_2$ | $\Leftrightarrow s \models \phi_1 \text{ and } s \models \phi_2$ |
| – $s \models \neg \phi$            | $\Leftrightarrow s \models \phi \text{ is false}$                |
| – $s \models P_{\sim p} [\psi]$    | $\Leftrightarrow \text{Prob}(s, \psi) \sim p$                    |
| – $s \models S_{\sim p} [\phi]$    | $\Leftrightarrow \sum_{s' \models \phi} \pi_s(s') \sim p$        |

Probability of,  
starting in state  $s$ ,  
satisfying the path  
formula  $\psi$

Probability of, starting in state  $s$ , being  
in state  $s'$  in the long run

# CSL semantics for CTMCs

- $\text{Prob}(s, \psi)$  is the probability, starting in state  $s$ , of satisfying the path formula  $\psi$

- $\text{Prob}(s, \psi) = \Pr_s \{ \omega \in \text{Path}_s \mid \omega \models \psi \}$

if  $\omega(0)$  is absorbing  $\omega(1)$  not defined

- Semantics of path formulae:

- for a path  $\omega$  of the CTMC:

- $\omega \models X \phi \iff \omega(1) \text{ is defined and } \omega(1) \models \phi$

- $\omega \models \phi_1 U^I \phi_2 \iff \exists t \in I. ( \omega@t \models \phi_2 \wedge \forall t' < t. \omega@t' \models \phi_1 )$

there exists a time instant in the **interval I** where  $\phi_2$  is true and  $\phi_1$  is true at all preceding time instants

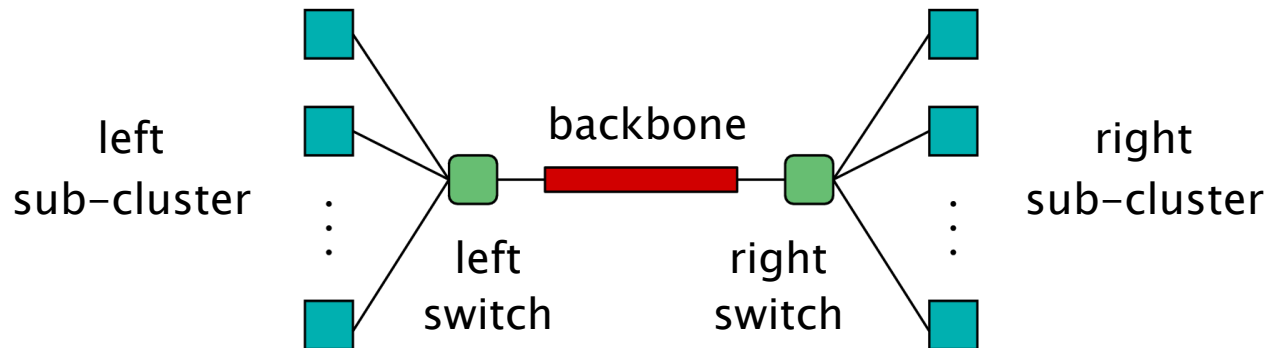
# More on CSL

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- Basic logical derivations:
  - false,  $\phi_1 \vee \phi_2$ ,  $\phi_1 \rightarrow \phi_2$
- Normal (unbounded) until is a special case
  - $\phi_1 \text{ U } \phi_2 \equiv \phi_1 \text{ U}^{[0, \infty)} \phi_2$
- Derived path formulae:
  - $F \phi \equiv \text{true U } \phi$ ,  $F^I \phi \equiv \text{true U}^I \phi$
  - $G \phi \equiv \neg(F \neg\phi)$ ,  $G^I \phi \equiv \neg(F^I \neg\phi)$
- Negate probabilities: ...
  - e.g.  $\neg P_{>p} [\psi] \equiv P_{\leq p} [\psi]$ ,  $\neg S_{\geq p} [\phi] \equiv S_{>p} [\phi]$
- Quantitative properties
  - of the form  $P_{=?} [\psi]$  and  $S_{=?} [\phi]$
  - where P/S is the outermost operator
  - experiments, patterns, trends, ...

# CSL example – Workstation cluster

- Case study: Cluster of workstations [HHK00]
  - two sub-clusters (N workstations in each cluster)
  - star topology with a central switch
  - components can break down, single repair unit



- **minimum QoS**: at least  $\frac{3}{4}$  of the workstations operational and connected via switches
- **premium QoS**: all workstations operational and connected via switches



# CSL example – Workstation cluster

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- $S_{=?} [\text{minimum}]$ 
  - the probability in the long run of having minimum QoS
- $P_{=?} [F^{[t,t]} \text{minimum}]$ 
  - the (transient) probability at time instant  $t$  of minimum QoS
- $P_{<0.05} [F^{[0,10]} \neg \text{minimum}]$ 
  - the probability that the QoS drops below minimum within 10 hours is less than 0.05
- $\neg \text{minimum} \rightarrow P_{<0.1} [F^{[0,2]} \neg \text{minimum}]$ 
  - when facing insufficient QoS, the chance of facing the same problem after 2 hours is less than 0.1

# CSL example – Workstation cluster

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- $\text{minimum} \rightarrow P_{>0.8} [ \text{minimum } U^{[0,t]} \text{ premium} ]$ 
  - the probability of going from minimum to premium QoS within  $t$  hours without violating minimum QoS is at least 0.8
- $P_{=?} [ \neg \text{minimum } U^{[t,\infty)} \text{ minimum} ]$ 
  - the chance it takes more than  $t$  time units to recover from insufficient QoS
- $\neg r\_switch\_up \rightarrow P_{<0.1} [ \neg r\_switch\_up \cup \neg l\_switch\_up ]$ 
  - if the right switch has failed, the probability of the left switch failing before it is repaired is less than 0.1
- $P_{=?} [ F^{[2,\infty)} S_{>0.9} [ \text{minimum} ] ]$ 
  - the probability of it taking more than 2 hours to get to a state from which the long-run probability of minimum QoS is  $>0.9$

# Summing up...

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- Transient probabilities (time instant  $t$ )
  - computation with uniformisation: efficient iterative method
- Steady-state (long-run) probabilities
  - like DTMCs
  - requires graph analysis
  - irreducible case: solve linear equation system
  - reducible case: steady-state for sub-CTMCs + reachability
- CSL: Continuous Stochastic Logic
  - extension of PCTL for properties of CTMCs