

Lecture 11

Counterexamples + Bisimulation

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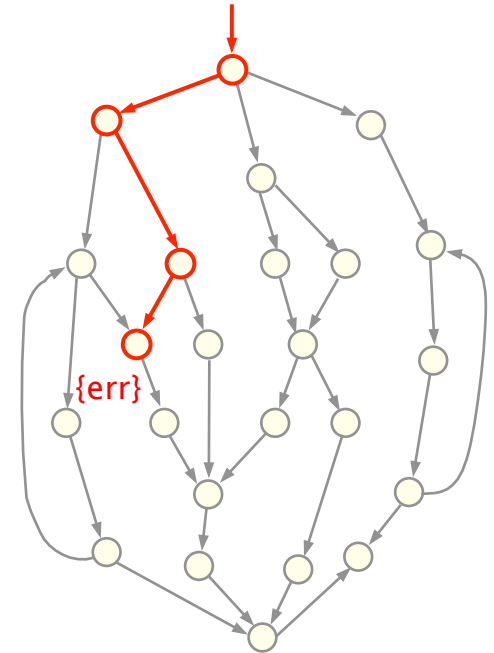
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Overview

- Counterexamples
 - non-probabilistic model checking
 - counterexamples for PCTL + DTMCs
 - computing smallest counterexamples
- Bisimulation
 - bisimulation equivalences: DTMCs, CTMCs
 - preservation of logics: PCTL, CSL
 - bisimulation minimisation

Non probabilistic counterexamples

- Counterexamples (for non-probabilistic model checking)
 - generated when model checking a (universal) property fails
 - trace through model illustrating why property does not hold
 - major advantage of the model checking approach
 - bug finding vs. verification
- Example:
 - CTL property $AG \neg \text{err}$
 - (or equivalently, $\neg EF \text{err}$)
 - (“an error state is never reached”)
 - counterexample is a finite trace to a state satisfying err
 - alternatively, this is a witness to the satisfaction of formula $EF \text{err}$



Counterexamples for DTMCs?

- PCTL example: $P_{<0.01} [F \text{ err }]$
 - “the probability of reaching an error state is less than 0.01”
 - what is a counterexample for $s \not\models P_{<0.01} [F \text{ err }]$?
 - not necessarily illustrated by a single trace to an **err** state
 - in fact, “counterexample” is a set of paths satisfying $F \text{ err}$ whose combined measure is greater than or equal to 0.01
- **Alternative approach to “debugging” seen so far:**
 - probabilistic model checker provides actual probabilities
 - e.g. queries of the form $P_{=?} [F \text{ err }]$
 - anomalous behaviour identified by examining trends
 - e.g. $P_{=?} [F^{\leq T} \text{ err }]$ for $T=0, \dots, 100$
- **This lecture: DTMC counterexamples in style of [HK07]**
 - also some work done on CTMC/MDP counterexamples

DTMC notation

- DTMC: $D = (S, s_{init}, P, L)$
- $\text{Path}(s)$ = set of all infinite paths starting in state s
- $\text{Pr}_s : \Sigma_{\text{Path}(s)} \rightarrow [0, 1]$ = probability measure over infinite paths
 - where $\Sigma_{\text{Path}(s)}$ is the σ -algebra on $\text{Path}(s)$
 - defined in terms of probabilities for finite paths
- $\mathbf{P}_s(\omega)$ = probability for finite path $\omega = ss_1 \dots s_n$
 - $\mathbf{P}_s(s) = 1$
 - $\mathbf{P}_s(ss_1 \dots s_n) = \mathbf{P}(s, s_1) \cdot \mathbf{P}(s_1, s_2) \cdot \dots \cdot \mathbf{P}(s_{n-1}, s_n)$
 - extend notation to sets: $\mathbf{P}_s(C)$ for set of finite paths C
 - \mathbf{P}_s extends uniquely to Pr_s
- $\text{Path}(s, \psi) = \{ \omega \in \text{Path}(s) \mid \omega \models \psi \}$
 - $\text{Prob}(s, \psi) = \text{Pr}_s(\text{Path}(s, \psi))$
- $\text{Path}_{\text{fin}}(s, \psi)$ = set of finite paths from s satisfying ψ

Counterexamples for DTMCs

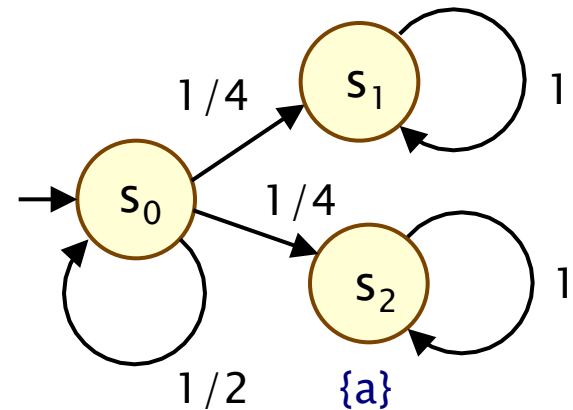
- Consider PCTL properties of the form:
 - $P_{\leq p} [\Phi_1 U^{\leq k} \Phi_2]$, where $k \in \mathbb{N} \cup \{\infty\}$
 - i.e. bounded or unbounded until formulae with closed upper probability bounds
- Refutation:
 - $s \not\models P_{\leq p} [\Phi_1 U^{\leq k} \Phi_2]$
 - $\Leftrightarrow \Pr_s(\text{Path}(s, \Phi_1 U^{\leq k} \Phi_2)) > p$
 - i.e. total probability mass of $\Phi_1 U^{\leq k} \Phi_2$ paths exceeds p
- Since the property is an until formula
 - this is evidenced by a set of finite paths

Counterexamples for DTMCs

- A counterexample for $P_{\leq p} [\Phi_1 U^{\leq k} \Phi_2]$ in state s is:
 - a set C of finite paths such that $C \subseteq \text{Path}_{\text{fin}}(s, \psi)$ and $P_s(C) > p$

- Example

- Consider the PCTL formula:
 - $P_{\leq 0.3} [F a]$
 - This is not satisfied in s_0
 - $\text{Prob}(s_0, F a) = 1/4 + 1/8 + 1/16 + \dots = 1/2$
 - A counterexample: $C = \{s_0 s_2, s_0 s_0 s_2\}$
 - $P_{s_0}(C) = 1/4 + (1/2)(1/4) = 3/8 = 0.375$



Finiteness of counterexamples

- There is always a finite counterexample for:

- $s \not\models P_{\leq p} [\phi_1 U^{\leq k} \phi_2]$

- On the other hand, consider this DTMC:

- and the PCTL formula:

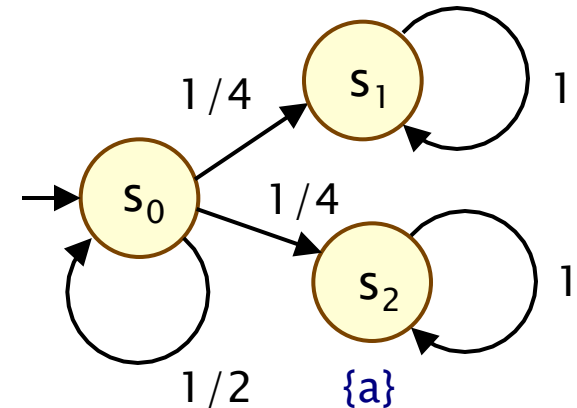
- $P_{<1/2} [F a]$

- $\text{Prob}(s_0, F a) = 1/4 + 1/8 + 1/16 + \dots$
 $= 1/2$

- $s_0 \not\models P_{<1/2} [F a]$

- counterexample would require infinite set of paths

- $\{ (s_0)^i s_2 \}_{i \in \mathbb{N}}$



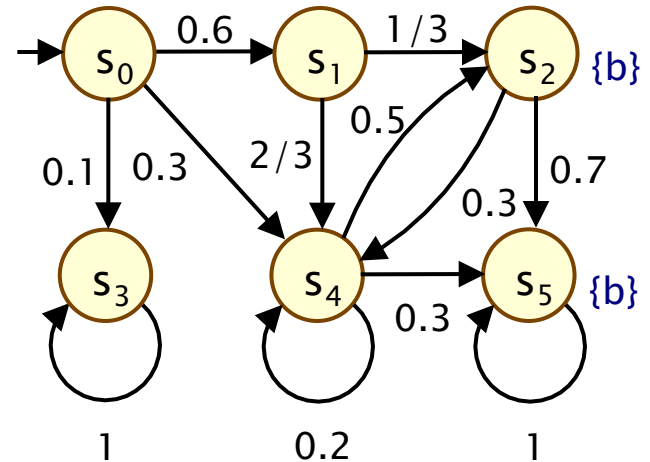
Counterexamples for DTMCs

- Aim: counterexamples should be succinct, comprehensible
- Set of all counterexamples:
 - $CX_p(s, \psi)$ = set of all counterexamples for $P_{\leq p} [\psi]$ in state s
- Minimal counterexample
 - counterexample C with $|C| \leq |C'|$ for all $C' \in CX_p(s, \psi)$
- “Smallest” counterexample
 - minimal counterexample C with $P(C) \geq P(C')$ for all minimal $C' \in CX_p(s, \psi)$
 - reduces to finding...
- Strongest (most probable) evidence
 - finite path ω in $\text{Path}_{\text{fin}}(s, \psi)$ such that $P(\omega) \geq P(\omega')$ for all $\omega' \in \text{Path}_{\text{fin}}(s, \psi)$
 - i.e. contributes most to violation of PCTL formula

Example

- PCTL formula: $P_{\leq 1/2} [F b]$

- $s_0 \not\models P_{\leq 1/2} [F b]$
- since $\text{Prob}(s_0, F b) = 0.9$



- Counterexamples:

- $C_1 = \{ s_0 s_1 s_2, s_0 s_1 s_4 s_2, s_0 s_1 s_4 s_5, s_0 s_4 s_2 \}$
 - $P_{s_0}(C_1) = 0.2 + 0.2 + 0.12 + 0.15 = 0.67$ (not minimal)
- $C_2 = \{ s_0 s_1 s_2, s_0 s_1 s_4 s_2, s_0 s_1 s_4 s_5 \}$
 - $P_{s_0}(C_2) = 0.2 + 0.2 + 0.12 = 0.52$ (not “smallest”)
- $C_3 = \{ s_0 s_1 s_2, s_0 s_1 s_4 s_2, s_0 s_4 s_2 \}$
 - $P_{s_0}(C_3) = 0.2 + 0.2 + 0.15 = 0.55$ (“smallest”)

Weighted digraphs

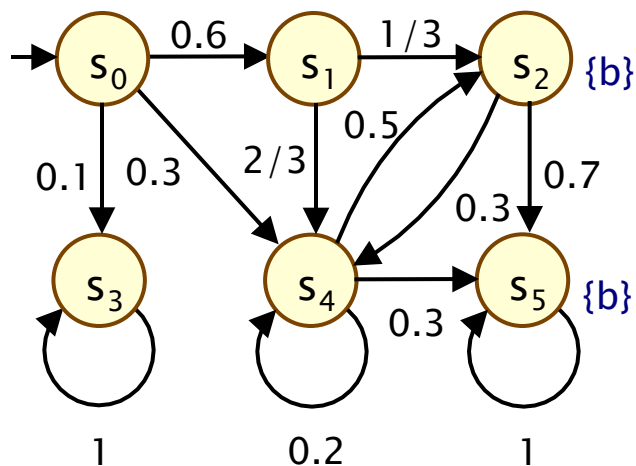
- A weighted directed graph is a tuple $G = (V, E, w)$ where:
 - V is a set of **vertices**
 - $E \subseteq V \times V$ is a set of **edges**
 - $w : E \rightarrow \mathbb{R}_{\geq 0}$ is a **weight function**
- **Finite path ω in G**
 - is a sequence of vertices $v_0 v_1 v_2 \dots v_n$ such that $(v_i, v_{i+1}) \in E \ \forall i \geq 0$
 - the **distance** of $\omega = v_0 v_1 v_2 \dots v_n$ is: $\sum_{i=0 \dots n-1} w(v_i, v_{i+1})$
- **Shortest path problem**
 - given a weighted digraph, find a path between two vertices v_1 and v_2 with the **smallest distance**
 - i.e. a path ω s.t. $d(\omega) \leq d(\omega')$ for all other such paths ω'

Finding strongest evidences

- Reduction to graph problem...
- Step 1: Adapt the DTMC
 - make states satisfying $\neg\phi_1 \wedge \neg\phi_2$ absorbing
 - (i.e. replace all outgoing transitions with a single self-loop)
 - add an extra state t and replace all transitions from any ϕ_2 state with a single transition to t (with probability 1)
- Step 2: Convert new DTMC into a weighted digraph
 - for the (adapted) DTMC $D = (S, s_{init}, P, L)$:
 - corresponding graph is $G_D = (V, E, w)$ where:
 - $V = S$ and $E = \{ (s, s') \in S \times S \mid P(s, s') > 0 \}$
 - $w(s, s') = \log(1 / P(s, s'))$
- Key idea: for any two paths ω and ω' in D (and in G_D)
 - $P_s(\omega') \geq P_s(\omega)$ if and only if $d(\omega') \leq d(\omega)$

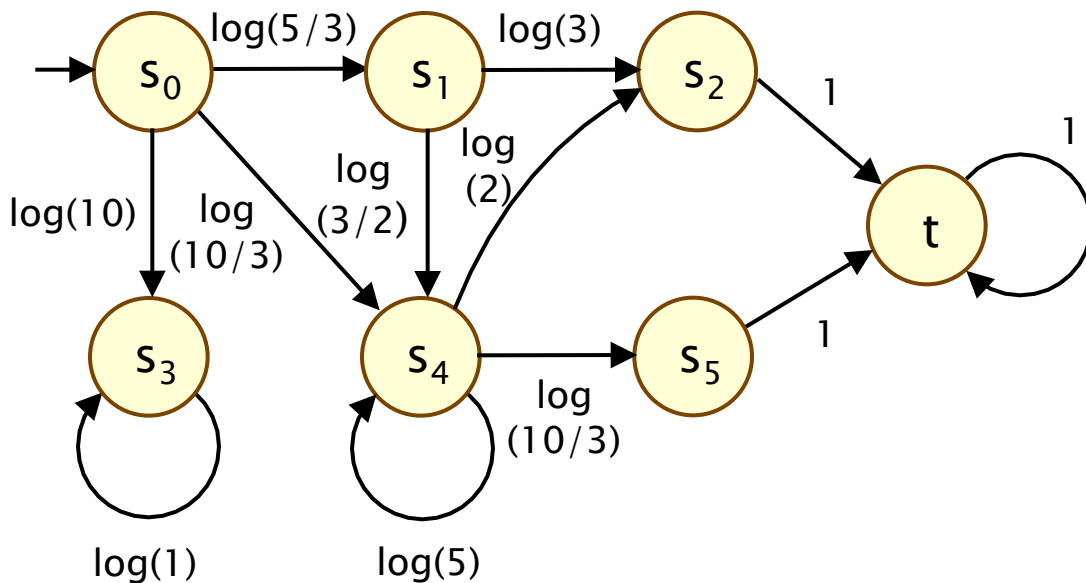
Example...

- PCTL formula: $P_{\leq 1/2} [F b]$



DTMC

weighted digraph



Finding strongest evidences

- To find strongest evidence in DTMC D
 - analyse corresponding digraph
- For unbounded until formula $P_{\leq p} [\Phi_1 U \Phi_2]$
 - solve shortest path problem in digraph (target t)
 - polynomial time algorithms exist
 - e.g. Dijkstra's algorithm can be implemented in $O(|E| + |V| \cdot \log |V|)$
- For bounded until formula $P_{\leq p} [\Phi_1 U^{\leq k} \Phi_2]$
 - solve special case of the constrained shortest path problem
 - also solvable in polynomial time
- Generation of smallest counterexamples
 - based on computation of k shortest paths
 - k can be computed on the fly

Other cases

- Lower bounds on probabilities
 - i.e. $s \not\models P_{\geq p} [\Phi_1 U^{\leq k} \Phi_2]$
 - negate until formula to reverse probability bound
 - solvable with BSCC computation + probabilistic reachability
 - for details, see [HK07]
- Continuous-time Markov chains
 - these techniques can be extended to CTMCs and CSL [HK07b]
 - naïve approach: apply DTMC techniques to uniformised DTMC
 - modifications required to get smaller counterexamples
 - another possibility: directed search based techniques [AHL05]

Bisimulation

- Identifies models with the same branching structure
 - i.e. the same stepwise behaviour
 - each model can simulate the actions of the other
 - guarantees that models satisfy many of the same properties
- Uses of bisimulation:
 - show equivalence between a model and its specification
 - state space reduction: bisimulation minimisation
- Formally, bisimulation is an equivalence relation over states
 - bisimilar states must have identical labelling and identical stepwise behaviour

Equivalence relations

- Let R be a relation over some set S
 - i.e. $R \subseteq S \times S$
 - we write $s_1 R s_2$ as shorthand for $(s_1, s_2) \in R$
- R is an equivalence relation iff:
 - R is **reflexive**, i.e. $s R s$
 - R is **symmetric**, i.e. if $s_1 R s_2$ then $s_2 R s_1$
 - R is **transitive**, i.e. if $s_1 R s_2$ and $s_2 R s_3$ then $s_1 R s_3$
- R partitions S :
 - **equivalence classes**: $[s]_R = \{ s' \in S \mid s' R s \}$
 - the **quotient** of S under R is denoted $S/R = \{ [s]_R \mid s \in S \}$

Bisimulation on DTMCs

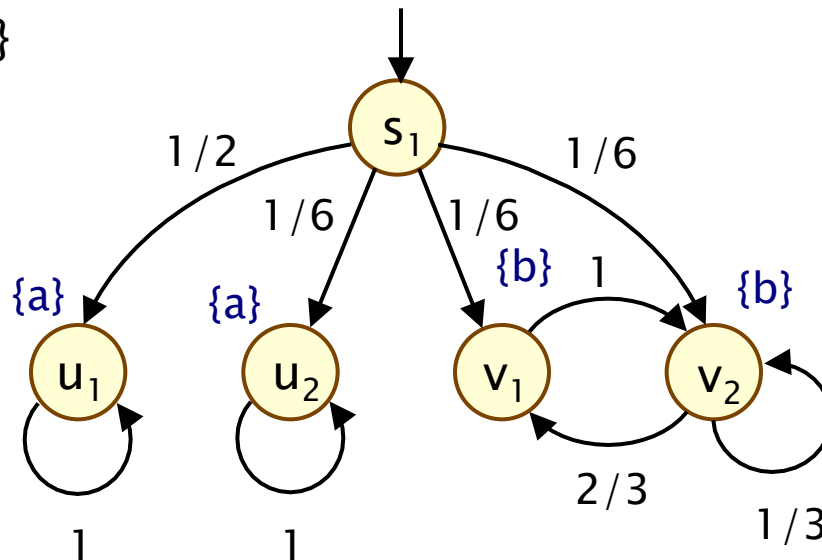
- Consider a DTMC $D = (S, s_{\text{init}}, P, L)$
- Some notation:
 - $P(s, T) = \sum_{s' \in T} P(s, s')$ for $T \subseteq S$
- An equivalence relation R on S is a **probabilistic bisimulation** on D if and only if for all $s_1 R s_2$:
 - $L(s_1) = L(s_2)$
 - $P(s_1, T) = P(s_2, T)$ for all $T \in S/R$ (i.e. for all equivalence classes of R)
- States s_1 and s_2 are **bisimulation-equivalent** (or **bisimilar**)
 - if there exists a probabilistic bisimulation R on D with $s_1 R s_2$
 - denoted $s_1 \sim s_2$

Simple example

- Bisimulation relation \sim
- Quotient of S under \sim
 - $\{ \{s_1\}, \{u_1, u_2\}, \{v_1, v_2\} \}$

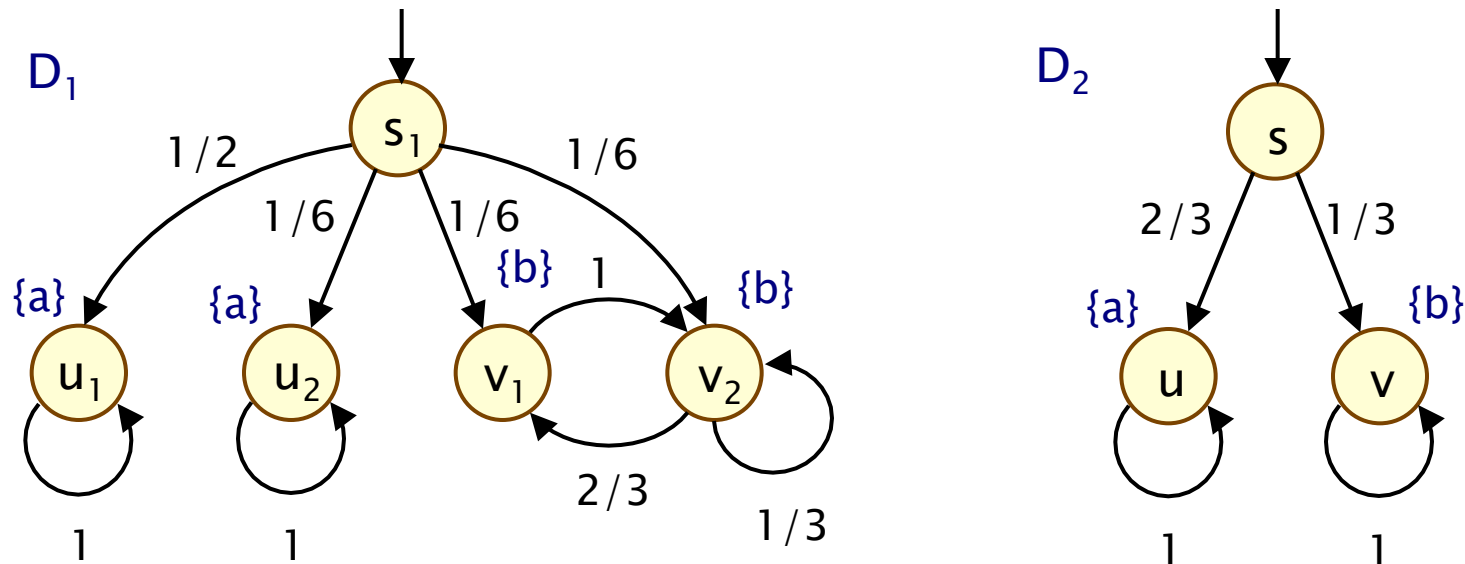
- Bisimilar states:

- $u_1 \sim u_2$
- $v_1 \sim v_2$



Bisimulation on DTMCs

- Bisimulation between DTMCs D_1 and D_2
 - $D_1 \sim D_2$ if they have bisimilar initial states
- Formally:
 - state labellings for D_1 and D_2 over same set of atomic prop.s
 - bisimulation relation is over disjoint union of D_1 and D_2

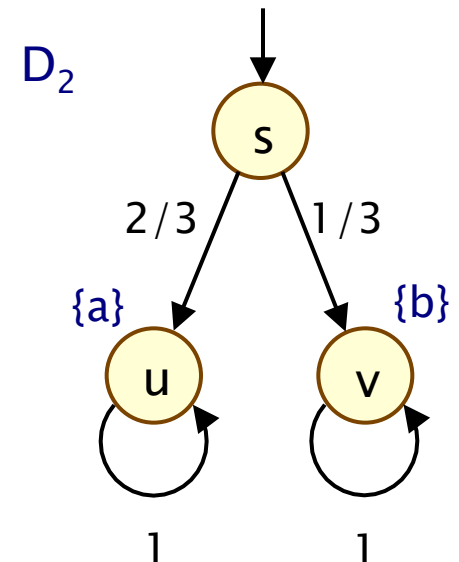
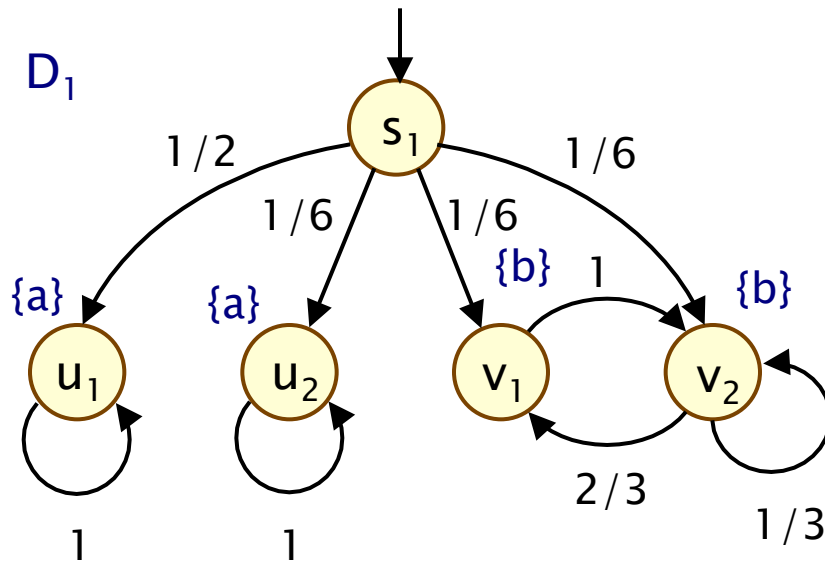


Simple example

- Bisimilar states:

- $u_1 \sim u_2 \sim u$
- $v_1 \sim v_2 \sim v$
- $s_1 \sim s$

Bisimilar DTMCs: $D_1 \sim D_2$



Quotient DTMC

- For a DTMC $D = (S, s_{\text{init}}, \mathbf{P}, L)$ and probabilistic bisimulation \sim

- Quotient DTMC is

- $D/\sim = (S', s'_{\text{init}}, \mathbf{P}', L')$

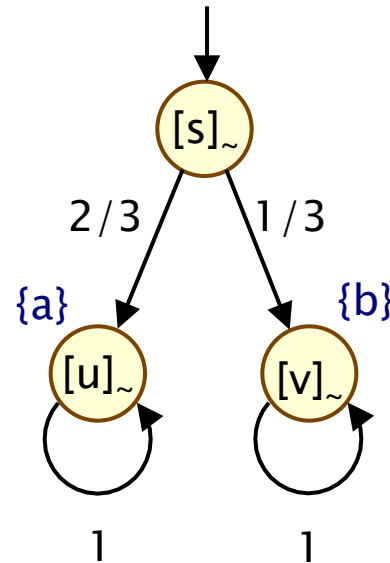
- where:

- $S' = S/\sim = \{ [s]_{\sim} \mid s \in S \}$

- $s'_{\text{init}} = [s_{\text{init}}]_{\sim}$

- $\mathbf{P}'([s]_{\sim}, [s']_{\sim}) = \mathbf{P}(s, [s']_{\sim})$

- $L'([s]_{\sim}) = L(s)$



well defined since
bisimulation ensures
 $\mathbf{P}(s, [s']_{\sim})$ same for all s in $[s]_{\sim}$

Bisimulation and PCTL

- Probabilistic bisimulation preserves all PCTL formulae
- For all states s and s' :

$$s \sim s' \\ \Leftrightarrow \\ \text{for all PCTL formulae } \phi, s \models \phi \text{ if and only if } s' \models \phi$$

- **Note also:**
 - every pair of non-bisimilar states can be distinguished with some PCTL formula
 - \sim is the coarsest relation with this property
 - in fact, bisimulation also preserves all PCTL* formulae

CTMC bisimulation

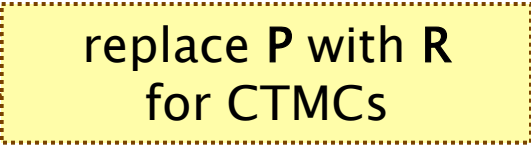
- Check equivalence of rates, not probabilities...
- An equivalence relation R on S is a probabilistic bisimulation on CTMC $C=(S,s_{init},R,L)$ if and only if for all $s_1 R s_2$:
 - $L(s_1) = L(s_2)$
 - $R(s_1, T) = R(s_2, T)$ for all classes T in S/R
- Alternatively, check:
 - $L(s_1) = L(s_2)$, $\mathbf{P}^{emb(C)}(s_1, T) = \mathbf{P}^{emb(C)}(s_2, T)$, $\mathbf{E}(s_1) = \mathbf{E}(s_2)$
- Bisimulation on CTMCs preserves CSL
 - (see [BHHK03] and also [DP03])

Bisimulation minimisation

- More efficient to perform PCTL/CSL model checking on the quotient DTMC/CTMC
 - assuming quotient model can be constructed efficiently
 - (see [KKZJ07] for experimental results on this)
- Bisimulation minimisation
 - algorithm to construct quotient model
 - based on partition refinement
 - repeated splitting of an initially coarse partition
 - final partition is coarsest bisimulation wrt. initial partition
 - (optimisations/variants possible by changing initial partition)
 - complexity: $O(|P| \cdot \log |S| + |AP| \cdot |S|)$ [DHS'03]
 - assuming suitable data structure used (splay trees)

Bisimulation minimisation

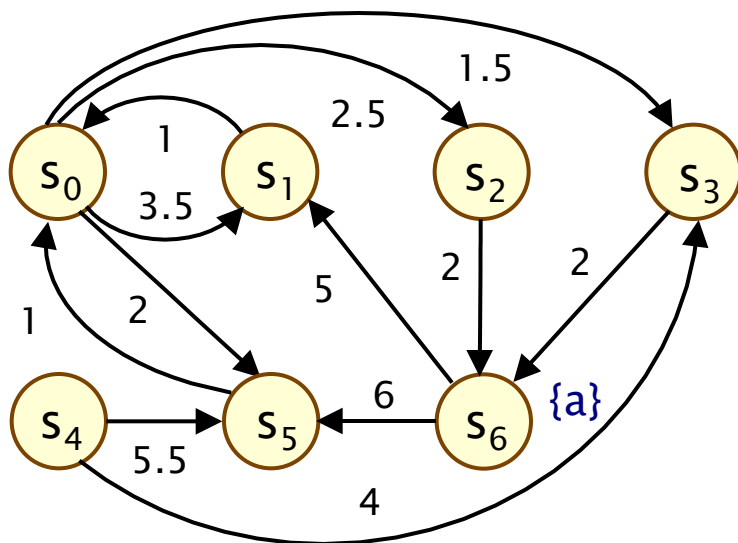
- 1. Start with **initial partition**
 - say $\Pi = \{ \{ s \in S \mid L(s) = \text{lab} \} \mid \text{lab} \in 2^{AP} \}$
- 2. Find a **splitter** $T \in \Pi$ for some block $B \in \Pi$
 - a splitter T is a block such that probability of going to T differs for some states in block B
 - i.e. $\exists s, s' \in B . P(s, T) \neq P(s', T)$
- 3. **Split** B into sub-blocks
 - such that $P(s, T)$ is the same for all states in each sub-block
- 4. **Repeat** steps 2/3 until no more splitters exist
 - i.e. no change to partition Π



replace P with R
for CTMCs

CTMC example

- Consider model checking $P_{\sim p} [F^{[0,t]} a]$ on this CTMC:



Minimisation:

$\Pi_0: B_1 = \{s_0, s_1, s_2, s_3, s_4, s_5\}, B_2 = \{s_6\}$

B_2 is a splitter for B_1

(since e.g. $R(s_1, B_2) = 0 \neq 2 = R(s_2, B_2)$)

$\Pi_1: B_1 = \{s_0, s_1, s_4, s_5\}, B_2 = \{s_6\}, B_3 = \{s_2, s_3\}$

B_3 is a splitter for B_1

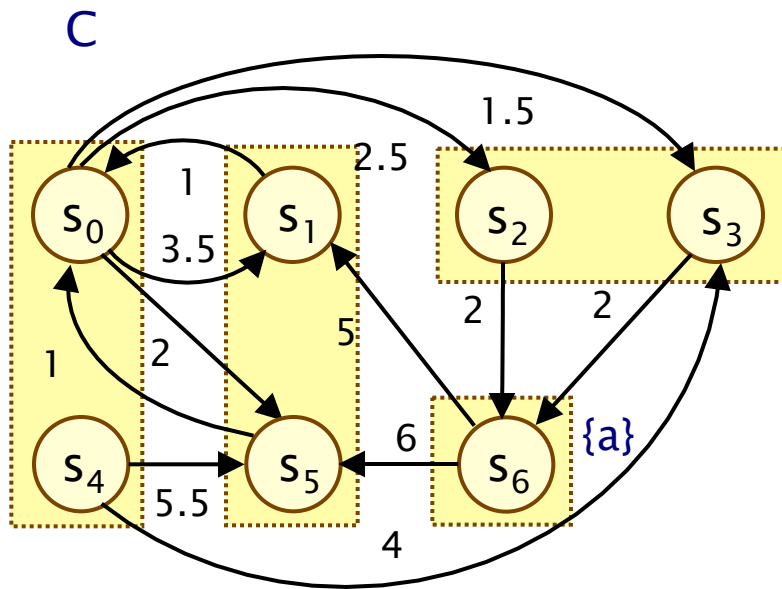
(since e.g. $R(s_1, B_3) = 0 \neq 4 = R(s_0, B_3)$)

$\Pi_2: B_1 = \{s_1, s_5\}, B_2 = \{s_6\}, B_3 = \{s_2, s_3\}, B_4 = \{s_0, s_4\}$

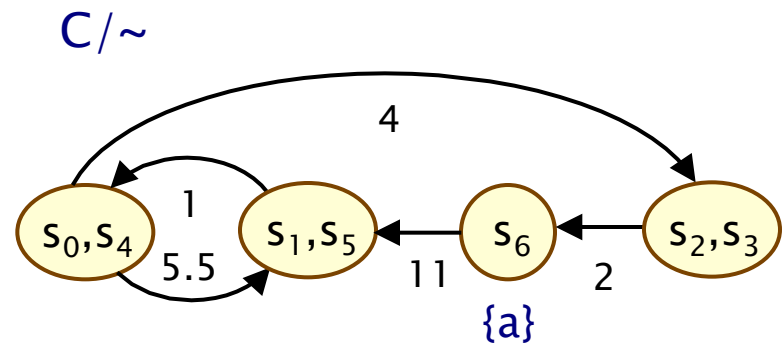
No more splitters...

$S/\sim = \{ \{s_1, s_5\}, \{s_6\}, \{s_2, s_3\}, \{s_0, s_4\} \}$

CTMC example...



$$S/\sim = \{ \{s_1, s_5\}, \{s_6\}, \{s_2, s_3\}, \{s_0, s_4\} \}$$



$$\text{Prob}^C(s_0, F^{[0,t]} a) = \text{Prob}^{C/\sim}(\{s_0, s_4\}, F^{[0,t]} a)$$

Summing up...

- Counterexamples

- essential ingredient of non-probabilistic model checking
- counterexamples for PCTL + DTMCs
 - finite set of paths showing $\not\models P_{\leq p} [\Phi_1 U^{\leq k} \Phi_2]$
- computing smallest counterexamples
 - reduction to well-known graph problems

- Bisimulation

- relates states/Markov chains with identical labelling and identical stepwise behaviour
- preserves PCTL, CSL, ...
- bisimulation minimisation: automated reduction to quotient model