



Probabilistic Model Checking

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Introduction

Probabilistic model checking

What is probabilistic model checking?

- Probabilistic model checking...
 - is a **formal verification** technique for modelling and analysing systems that exhibit **probabilistic** behaviour
- Formal verification...
 - is the application of rigorous, mathematics-based techniques to establish the correctness of computerised systems

Why formal verification?

- Errors in computerised systems can be costly...



Pentium chip (1994)

Bug found in FPU.
Intel (eventually) offers to replace faulty chips.
Estimated loss: \$475m



Ariane 5 (1996)

Self-destructs 37secs into maiden launch.
Cause: uncaught overflow exception.



Toyota Prius (2010)

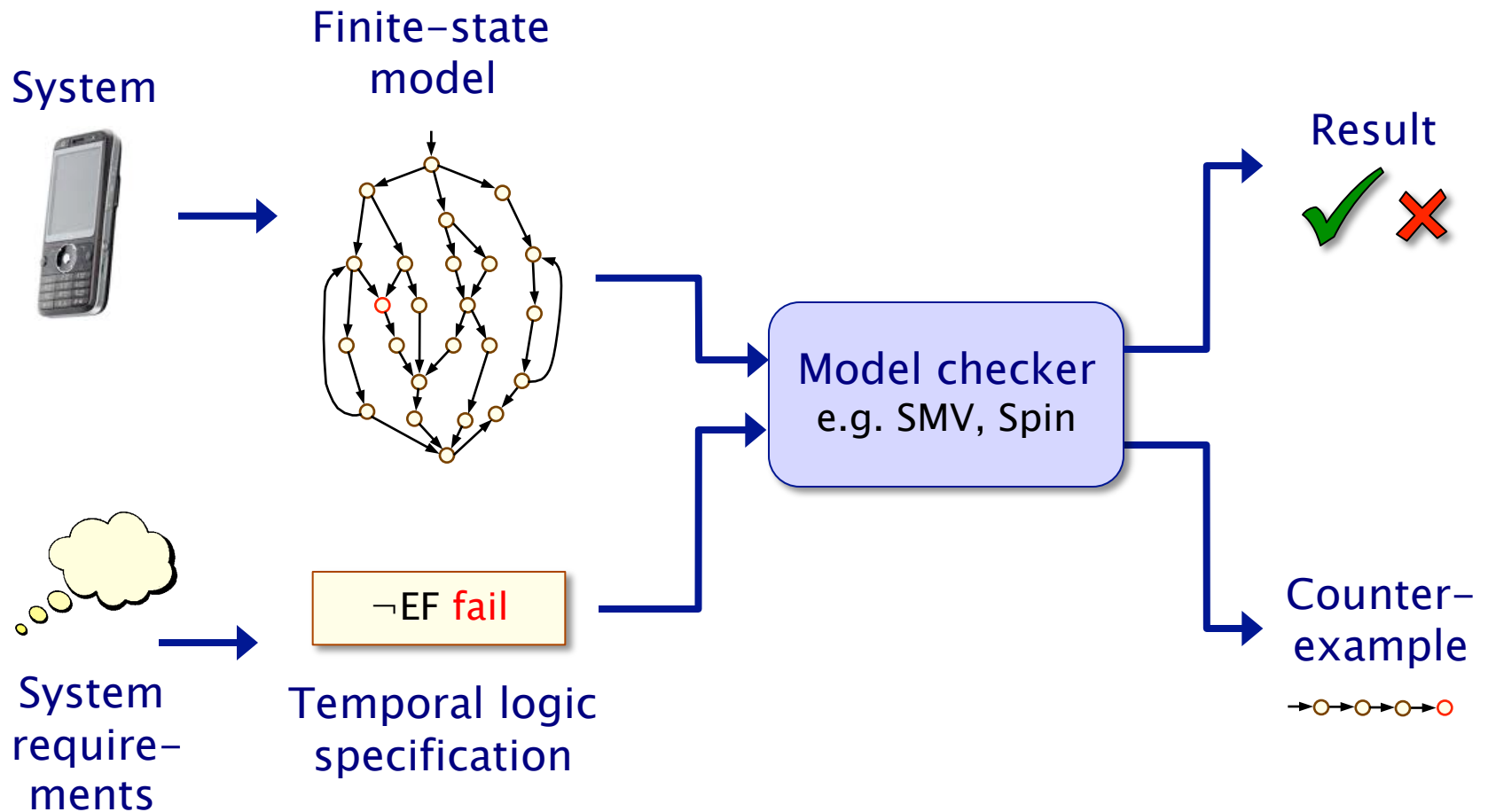
Software “glitch” found in anti-lock braking system.
185,000 cars recalled.

- Why verify?

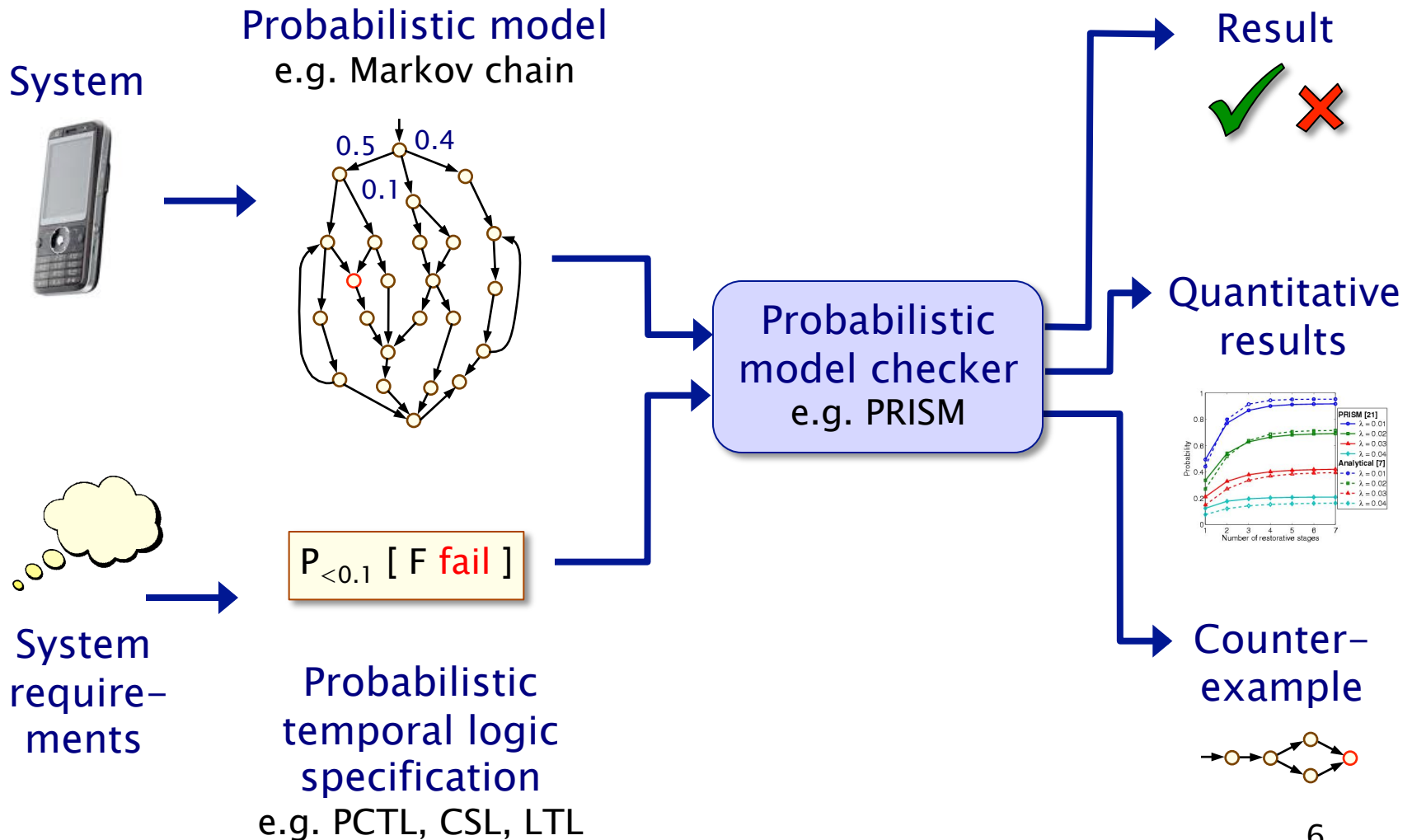
- “Testing can only show the presence of errors, not their absence.” [Edsger Dijkstra]



Model checking



Probabilistic model checking



Why probability?

- Some systems are inherently probabilistic...
- **Randomisation**, e.g. in distributed coordination algorithms
 - as a symmetry breaker, in gossip routing to reduce flooding
- **Examples: real-world protocols featuring randomisation:**
 - Randomised back-off schemes
 - CSMA protocol, 802.11 Wireless LAN
 - Random choice of waiting time
 - IEEE1394 Firewire (root contention), Bluetooth (device discovery)
 - Random choice over a set of possible addresses
 - IPv4 Zeroconf dynamic configuration (link-local addressing)
 - Randomised algorithms for anonymity, contract signing, ...

Why probability?

- Some systems are inherently probabilistic...
- **Randomisation**, e.g. in distributed coordination algorithms
 - as a symmetry breaker, in gossip routing to reduce flooding
- To model **uncertainty** and **performance**
 - to quantify rate of failures, express Quality of Service
- **Examples:**
 - computer networks, embedded systems
 - power management policies
 - nano-scale circuitry: reliability through defect-tolerance

Why probability?

- Some systems are inherently probabilistic...
- **Randomisation**, e.g. in distributed coordination algorithms
 - as a symmetry breaker, in gossip routing to reduce flooding
- To model **uncertainty and performance**
 - to quantify rate of failures, express Quality of Service
- To model **biological processes**
 - reactions occurring between large numbers of molecules are naturally modelled in a stochastic fashion

Verifying probabilistic systems

- We are not just interested in correctness
- We want to be able to quantify:
 - security, privacy, trust, anonymity, fairness
 - safety, reliability, performance, dependability
 - resource usage, e.g. battery life
 - and much more...
- **Quantitative**, as well as qualitative requirements:
 - how reliable is my car's Bluetooth network?
 - how efficient is my phone's power management policy?
 - is my bank's web-service secure?
 - what is the expected long-run percentage of protein X?

Probabilistic models

	Fully probabilistic	Nondeterministic
Discrete time	Discrete-time Markov chains (DTMCs)	Markov decision processes (MDPs) (probabilistic automata)
Continuous time	Continuous-time Markov chains (CTMCs)	CTMDPs/IMCs
		Probabilistic timed automata (PTAs)

Course overview

- 5 lectures: Mon–Fri, 11am–12.30pm
 - Introduction
 - 1 – Discrete time Markov chains
 - 2 – Markov decision processes
 - 3 – Continuous–time Markov chains
 - 4 – Probabilistic model checking in practice
 - 5 – Probabilistic timed automata
- Course materials available here:
 - <http://www.prismmodelchecker.org/lectures/esslli10/>
 - lecture slides, reference list



Part 1

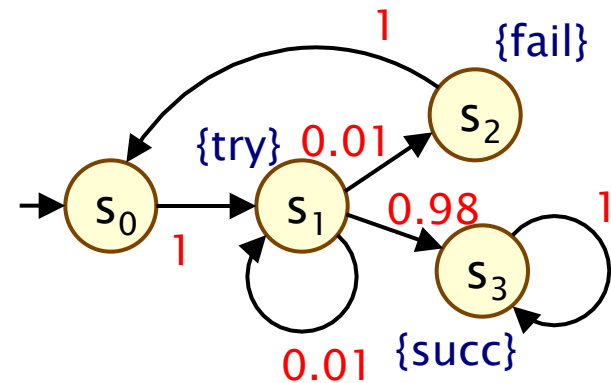
Discrete-time Markov chains

Overview (Part 1)

- Discrete-time Markov chains (DTMCs)
- PCTL: A temporal logic for DTMCs
- PCTL model checking
- LTL model checking
- Costs and rewards

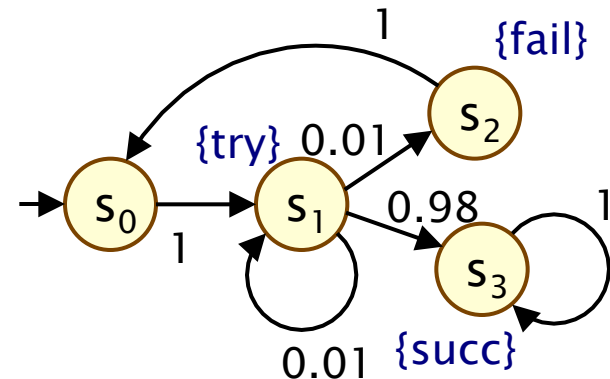
Discrete-time Markov chains

- Discrete-time Markov chains (DTMCs)
 - state-transition systems augmented with probabilities
- States
 - **discrete set of states** representing possible configurations of the system being modelled
- Transitions
 - transitions between states occur in **discrete time-steps**
- Probabilities
 - probability of making transitions between states is given by **discrete probability distributions**



Discrete-time Markov chains

- Formally, a DTMC D is a tuple $(S, s_{\text{init}}, P, L)$ where:
 - S is a finite set of states (“state space”)
 - $s_{\text{init}} \in S$ is the initial state
 - $P : S \times S \rightarrow [0,1]$ is the **transition probability matrix** where $\sum_{s' \in S} P(s, s') = 1$ for all $s \in S$
 - $L : S \rightarrow 2^{\text{AP}}$ is function labelling states with atomic propositions
- Note: no deadlock states**
 - i.e. every state has at least one outgoing transition
 - can add self loops to represent final/terminating states

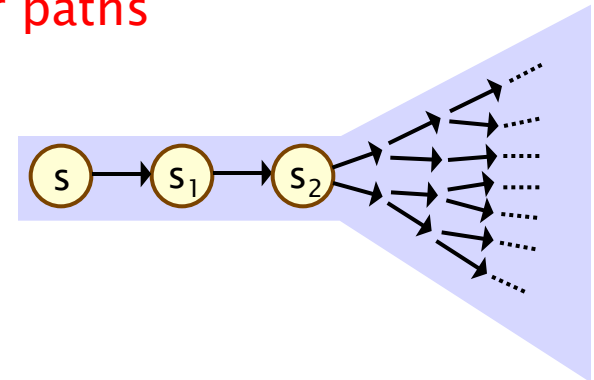


DTMCs: An alternative definition

- **Alternative definition: a DTMC is:**
 - a family of **random variables** $\{ X(k) \mid k=0,1,2,\dots \}$
 - $X(k)$ are observations at discrete time-steps
 - i.e. $X(k)$ is the state of the system at time-step k
- **Memorylessness (Markov property)**
 - $\Pr(X(k)=s_k \mid X(k-1)=s_{k-1}, \dots, X(0)=s_0)$
= $\Pr(X(k)=s_k \mid X(k-1)=s_{k-1})$
- **We consider homogenous DTMCs**
 - transition probabilities are **independent of time**
 - $P(s_{k-1},s_k) = \Pr(X(k)=s_k \mid X(k-1)=s_{k-1})$

Paths and probabilities

- A (finite or infinite) path through a DTMC
 - is a sequence of states $s_0s_1s_2s_3\dots$ such that $P(s_i, s_{i+1}) > 0 \forall i$
 - represents an **execution** (i.e. one possible behaviour) of the system which the DTMC is modelling
- To reason (quantitatively) about this system
 - need to define a **probability space over paths**
- Intuitively:
 - sample space: $\text{Path}(s) =$ set of all infinite paths from a state s
 - events: sets of infinite paths from s
 - basic events: **cylinder sets** (or “cones”)
 - cylinder set $C(\omega)$, for a finite path ω
= set of **infinite paths with the common finite prefix ω**
 - for example: $C(ss_1s_2)$



Probability spaces

- Let Ω be an arbitrary non-empty set
- A **σ -algebra** (or σ -field) on Ω is a family Σ of subsets of Ω closed under complementation and countable union, i.e.:
 - if $A \in \Sigma$, the complement $\Omega \setminus A$ is in Σ
 - if $A_i \in \Sigma$ for $i \in \mathbb{N}$, the union $\cup_i A_i$ is in Σ
 - the empty set \emptyset is in Σ
- **Theorem:** For any family F of subsets of Ω , there exists a unique smallest σ -algebra on Ω containing F
- **Probability space $(\Omega, \Sigma, \text{Pr})$**
 - Ω is the sample space
 - Σ is the set of events: σ -algebra on Ω
 - $\text{Pr} : \Sigma \rightarrow [0,1]$ is the probability measure:
 $\text{Pr}(\Omega) = 1$ and $\text{Pr}(\cup_i A_i) = \sum_i \text{Pr}(A_i)$ for countable disjoint A_i

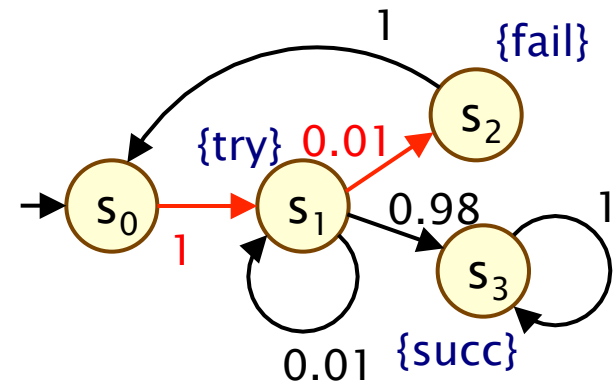
Probability space over paths

- Sample space $\Omega = \text{Path}(s)$
set of infinite paths with initial state s
- Event set $\Sigma_{\text{Path}(s)}$
 - the **cylinder set** $C(\omega) = \{ \omega' \in \text{Path}(s) \mid \omega \text{ is prefix of } \omega' \}$
 - $\Sigma_{\text{Path}(s)}$ is the **least σ -algebra** on $\text{Path}(s)$ containing $C(\omega)$ for all finite paths ω starting in s
- Probability measure \Pr_s
 - define probability $P_s(\omega)$ for finite path $\omega = ss_1 \dots s_n$ as:
 - $P_s(\omega) = 1$ if ω has length one (i.e. $\omega = s$)
 - $P_s(\omega) = P(s, s_1) \cdot \dots \cdot P(s_{n-1}, s_n)$ otherwise
 - define $\Pr_s(C(\omega)) = P_s(\omega)$ for all finite paths ω
 - \Pr_s extends **uniquely** to a probability measure $\Pr_s: \Sigma_{\text{Path}(s)} \rightarrow [0, 1]$
- See [KSK76] for further details

Probability space – Example

- Paths where sending fails the first time

- $\omega = s_0s_1s_2$
- $C(\omega) =$ all paths starting $s_0s_1s_2\dots$
- $P_{s_0}(\omega) = P(s_0, s_1) \cdot P(s_1, s_2)$
 $= 1 \cdot 0.01 = 0.01$
- $\Pr_{s_0}(C(\omega)) = P_{s_0}(\omega) = 0.01$



- Paths which are eventually successful and with no failures

- $C(s_0s_1s_3) \cup C(s_0s_1s_1s_3) \cup C(s_0s_1s_1s_1s_3) \cup \dots$
- $\Pr_{s_0}(C(s_0s_1s_3) \cup C(s_0s_1s_1s_3) \cup C(s_0s_1s_1s_1s_3) \cup \dots)$
 $= P_{s_0}(s_0s_1s_3) + P_{s_0}(s_0s_1s_1s_3) + P_{s_0}(s_0s_1s_1s_1s_3) + \dots$
 $= 1 \cdot 0.98 + 1 \cdot 0.01 \cdot 0.98 + 1 \cdot 0.01 \cdot 0.01 \cdot 0.98 + \dots$
 $= 0.9898989898\dots$
 $= 98/99$

Overview (Part 1)

- Discrete-time Markov chains (DTMCs)
- **PCTL: A temporal logic for DTMCs**
- PCTL model checking
- LTL model checking
- Costs and rewards

PCTL

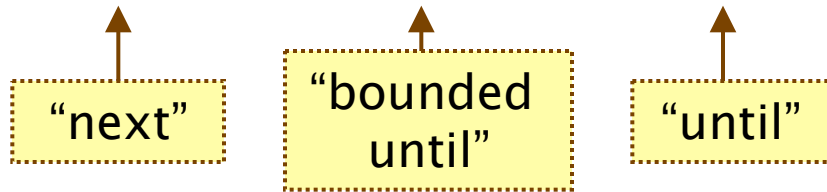
- Temporal logic for describing properties of DTMCs
 - PCTL = Probabilistic Computation Tree Logic [HJ94]
 - essentially the same as the logic pCTL of [ASB+95]
- Extension of (non-probabilistic) temporal logic CTL
 - key addition is **probabilistic operator P**
 - quantitative extension of CTL's A and E operators
- Example
 - $\text{send} \rightarrow P_{\geq 0.95} [\text{true } U^{\leq 10} \text{deliver}]$
 - “if a message is sent, then the probability of it being delivered within 10 steps is at least 0.95”

PCTL syntax

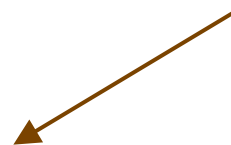
- PCTL syntax:

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

– $\psi ::= X\phi \mid \phi U^{\leq k} \phi \mid \phi U \phi$ (path formulas)



ψ is true with probability $\sim p$



– where a is an atomic proposition, used to identify states of interest, $p \in [0,1]$ is a probability, $\sim \in \{<, >, \leq, \geq\}$, $k \in \mathbb{N}$

- A PCTL formula is always a state formula

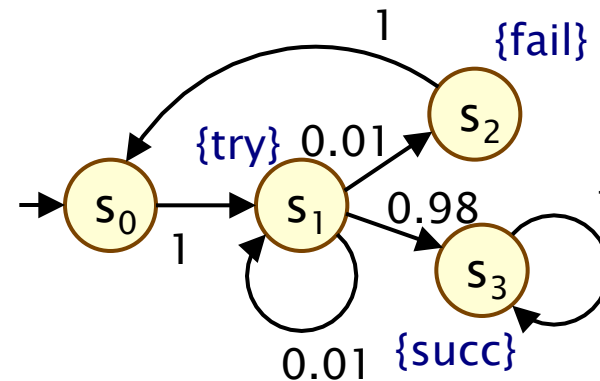
– path formulas only occur inside the P operator

PCTL semantics for DTMCs

- PCTL formulas interpreted over states of a DTMC
 - $s \models \phi$ denotes ϕ is “true in state s ” or “satisfied in state s ”
- Semantics of (non-probabilistic) state formulas:
 - for a state s of the DTMC (S, s_{init}, P, L) :
 - $s \models a \iff a \in L(s)$
 - $s \models \phi_1 \wedge \phi_2 \iff s \models \phi_1$ and $s \models \phi_2$
 - $s \models \neg\phi \iff s \models \phi$ is false

- Examples

- $s_3 \models \text{succ}$
- $s_1 \models \text{try} \wedge \neg\text{fail}$



PCTL semantics for DTMCs

- Semantics of path formulas:

- for a path $\omega = s_0s_1s_2\dots$ in the DTMC:

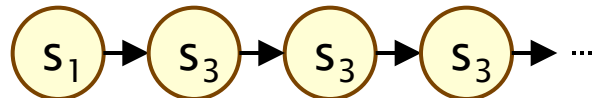
- $\omega \models X \phi \iff s_1 \models \phi$

- $\omega \models \phi_1 U^{\leq k} \phi_2 \iff \exists i \leq k$ such that $s_i \models \phi_2$ and $\forall j < i, s_j \models \phi_1$

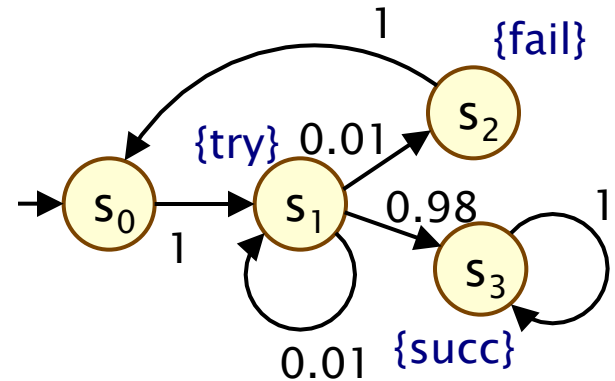
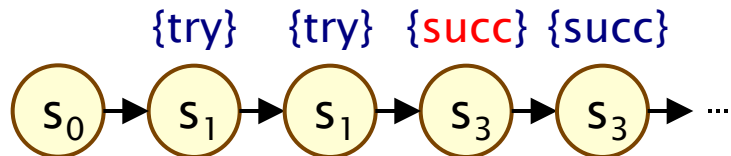
- $\omega \models \phi_1 U \phi_2 \iff \exists k \geq 0$ such that $\omega \models \phi_1 U^{\leq k} \phi_2$

- Some examples of satisfying paths:

- $X \text{succ}$ {try} {succ} {succ} {succ}



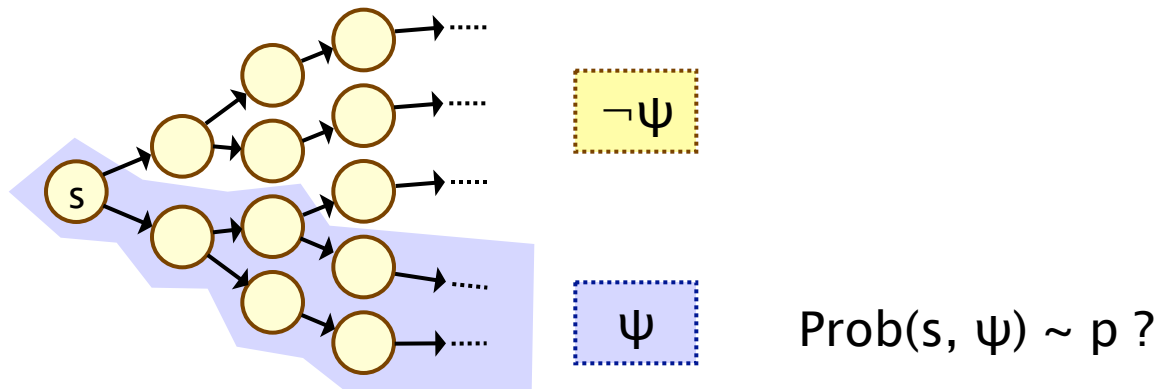
- $\neg \text{fail} U \text{succ}$



PCTL semantics for DTMCs

- Semantics of the probabilistic operator P

- informal definition: $s \models P_{\sim p} [\psi]$ means that “**the probability, from state s , that ψ is true for an outgoing path satisfies $\sim p$** ”
- example: $s \models P_{<0.25} [X \text{ fail}] \Leftrightarrow$ “the probability of atomic proposition fail being true in the next state of outgoing paths from s is less than 0.25”
- formally: $s \models P_{\sim p} [\psi] \Leftrightarrow \text{Prob}(s, \psi) \sim p$
- where: $\text{Prob}(s, \psi) = \Pr_s \{ \omega \in \text{Path}(s) \mid \omega \models \psi \}$
- (sets of paths satisfying ψ are always measurable [Var85])



More PCTL...

- Usual temporal logic equivalences:

- $\text{false} \equiv \neg \text{true}$

(false)

- $\phi_1 \vee \phi_2 \equiv \neg(\neg\phi_1 \wedge \neg\phi_2)$

(disjunction)

- $\phi_1 \rightarrow \phi_2 \equiv \neg\phi_1 \vee \phi_2$

(implication)

- $F\phi \equiv \diamond\phi \equiv \text{true} \cup \phi$

(eventually, “future”)

- $G\phi \equiv \square\phi \equiv \neg(F\neg\phi)$

(always, “globally”)

- bounded variants: $F^{\leq k}\phi$, $G^{\leq k}\phi$

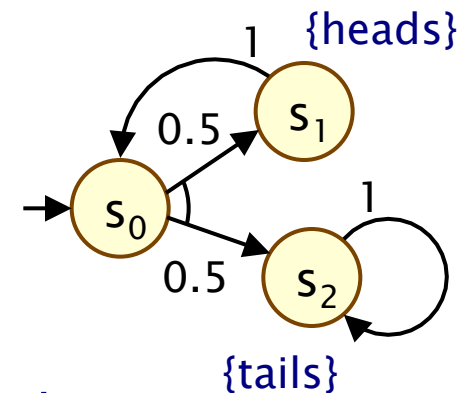
- Negation and probabilities

- e.g. $\neg P_{>p}[\phi_1 \cup \phi_2] \equiv P_{\leq p}[\phi_1 \cup \phi_2]$

- e.g. $P_{>p}[G\phi] \equiv P_{<1-p}[F\neg\phi]$

Qualitative vs. quantitative properties

- P operator of PCTL can be seen as a **quantitative** analogue of the CTL operators A (for all) and E (there exists)
- A PCTL property $P_{\sim p} [\psi]$ is...
 - **qualitative** when p is either 0 or 1
 - **quantitative** when p is in the range (0,1)
- $P_{>0} [F \phi]$ is identical to $EF \phi$
 - there exists a finite path to a ϕ -state
- $P_{\geq 1} [F \phi]$ is (similar to but) weaker than $AF \phi$
 - e.g. **AF “tails”** (CTL) \neq **$P_{\geq 1} [F \text{“tails”}]$** (PCTL)

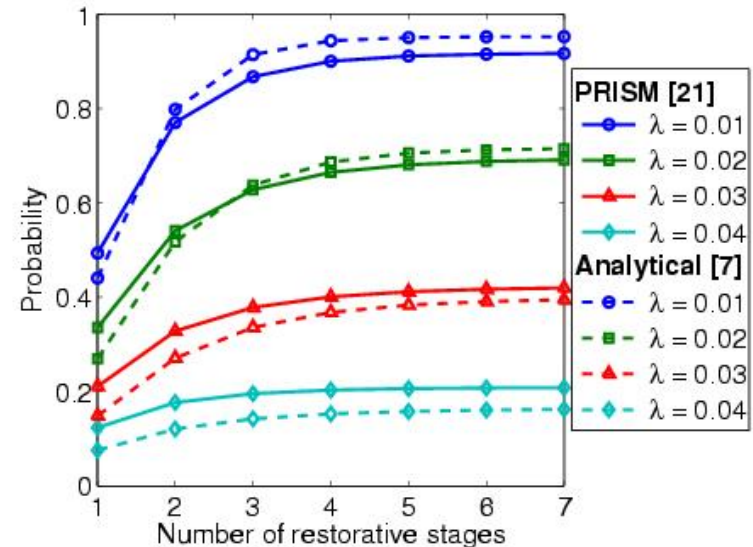


Quantitative properties

- Consider a PCTL formula $P_{\sim p} [\psi]$
 - if the probability is **unknown**, how to choose the bound p ?
- When the outermost operator of a PTCL formula is P
 - we allow the form $P_{=?} [\psi]$
 - “**what is the probability that path formula ψ is true?**”
- Model checking is no harder: compute the values anyway
- Useful to spot patterns, trends

- **Example**

- $P_{=?} [F \text{ err}/\text{total} > 0.1]$
- “what is the probability that 10% of the NAND gate outputs are erroneous?”



Some real PCTL examples

- **NAND multiplexing system**

- $P_{=?} [F \text{ err/total} > 0.1]$
- “what is the probability that 10% of the NAND gate outputs are erroneous?”

reliability

- **Bluetooth wireless communication protocol**

- $P_{=?} [F^{\leq t} \text{ reply_count} = k]$
- “what is the probability that the sender has received k acknowledgements within t clock-ticks?”

performance

- **Security: EGL contract signing protocol**

- $P_{=?} [F (\text{pairs_a} = 0 \ \& \ \text{pairs_b} > 0)]$
- “what is the probability that the party B gains an unfair advantage during the execution of the protocol?”

fairness

Overview (Part 1)

- Discrete-time Markov chains (DTMCs)
- PCTL: A temporal logic for DTMCs
- **PCTL model checking**
- LTL model checking
- Costs and rewards

PCTL model checking for DTMCs

- Algorithm for PCTL model checking [CY88,HJ94,CY95]
 - inputs: DTMC $D=(S,s_{init},P,L)$, PCTL formula ϕ
 - output: $Sat(\phi) = \{ s \in S \mid s \models \phi \}$ = set of states satisfying ϕ
- What does it mean for a DTMC D to satisfy a formula ϕ ?
 - sometimes, want to check that $s \models \phi \ \forall s \in S$, i.e. $Sat(\phi) = S$
 - sometimes, just want to know if $s_{init} \models \phi$, i.e. if $s_{init} \in Sat(\phi)$
- Sometimes, focus on **quantitative** results
 - e.g. compute result of $P=?$ [F error]
 - e.g. compute result of $P=?$ [$F^{\leq k}$ error] for $0 \leq k \leq 100$

PCTL model checking for DTMCs

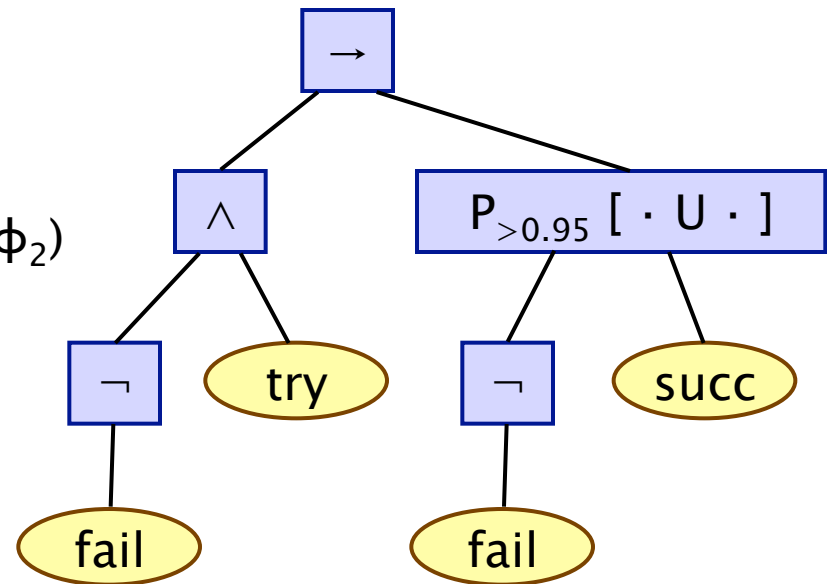
- Basic algorithm proceeds by induction on parse tree of ϕ
 - example: $\phi = (\neg\text{fail} \wedge \text{try}) \rightarrow P_{>0.95} [\neg\text{fail} \cup \text{succ}]$

- For the non-probabilistic operators:

- $\text{Sat}(\text{true}) = S$
- $\text{Sat}(a) = \{ s \in S \mid a \in L(s) \}$
- $\text{Sat}(\neg\phi) = S \setminus \text{Sat}(\phi)$
- $\text{Sat}(\phi_1 \wedge \phi_2) = \text{Sat}(\phi_1) \cap \text{Sat}(\phi_2)$

- For the $P_{\sim p} [\psi]$ operator

- need to compute the probabilities $\text{Prob}(s, \psi)$ for all states $s \in S$
- focus here on “until” case: $\psi = \phi_1 \cup \phi_2$



PCTL until for DTMCs

- Computation of probabilities $\text{Prob}(s, \phi_1 \cup \phi_2)$ for all $s \in S$
- First, identify all states where the **probability** is **1** or **0**
 - $S^{\text{yes}} = \text{Sat}(P_{\geq 1} [\phi_1 \cup \phi_2])$
 - $S^{\text{no}} = \text{Sat}(P_{\leq 0} [\phi_1 \cup \phi_2])$
- Then solve linear equation system for remaining states
- We refer to the first phase as “**precomputation**”
 - two algorithms: Prob0 (for S^{no}) and Prob1 (for S^{yes})
 - algorithms work on underlying graph (probabilities irrelevant)
- Important for several reasons
 - reduces the set of states for which probabilities must be computed numerically (which is more expensive)
 - gives **exact results** for the states in S^{yes} and S^{no} (no round-off)
 - for $P_{\sim p}[\cdot]$ where p is 0 or 1, no further computation required

PCTL until – Linear equations

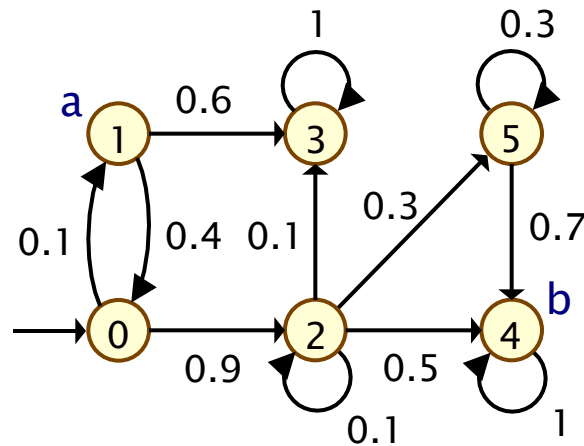
- Probabilities $\text{Prob}(s, \phi_1 \cup \phi_2)$ can now be obtained as the unique solution of the following set of **linear equations**:

$$\text{Prob}(s, \phi_1 \cup \phi_2) = \begin{cases} 1 & \text{if } s \in S^{\text{yes}} \\ 0 & \text{if } s \in S^{\text{no}} \\ \sum_{s' \in S} P(s, s') \cdot \text{Prob}(s', \phi_1 \cup \phi_2) & \text{otherwise} \end{cases}$$

- can be reduced to a system in $|S^?|$ unknowns instead of $|S|$ where $S^? = S \setminus (S^{\text{yes}} \cup S^{\text{no}})$
- This can be solved with (a variety of) standard techniques
 - direct methods, e.g. Gaussian elimination
 - iterative methods, e.g. Jacobi, Gauss–Seidel, ... (preferred in practice due to scalability)

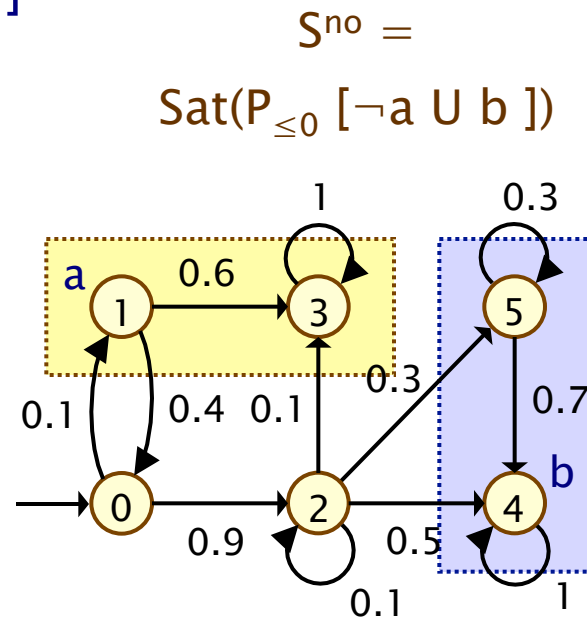
PCTL until – Example

- Example: $P_{>0.8} [\neg a \text{ U } b]$



PCTL until – Example

- Example: $P_{>0.8} [\neg a \text{ U } b]$



$S^{\text{yes}} =$
 $\text{Sat}(P_{\geq 1} [\neg a \text{ U } b])$

PCTL until – Example

- Example: $P_{>0.8} [\neg a \text{ U } b]$

- Let $x_s = \text{Prob}(s, \neg a \text{ U } b)$

- Solve:

$$x_4 = x_5 = 1$$

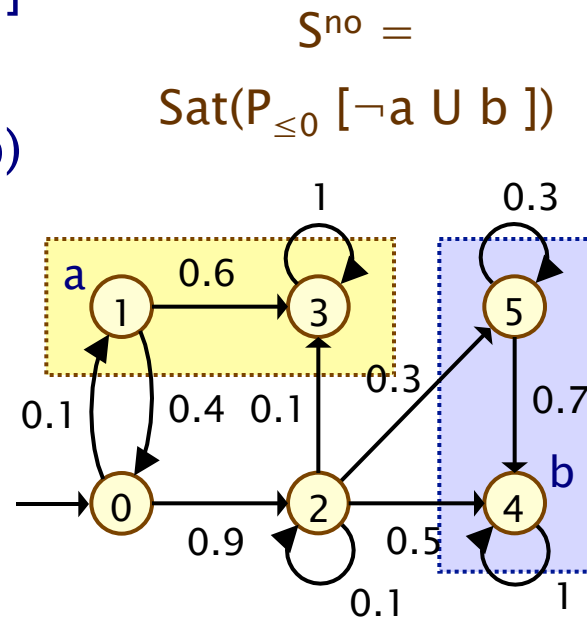
$$x_1 = x_3 = 0$$

$$x_0 = 0.1x_1 + 0.9x_2 = 0.8$$

$$x_2 = 0.1x_2 + 0.1x_3 + 0.3x_5 + 0.5x_4 = 8/9$$

$$\text{Prob}(\neg a \text{ U } b) = \underline{x} = [0.8, 0, 8/9, 0, 1, 1]$$

$$\text{Sat}(P_{>0.8} [\neg a \text{ U } b]) = \{s_2, s_4, s_5\}$$



$S^{\text{yes}} =$

$\text{Sat}(P_{\geq 1} [\neg a \text{ U } b])$

PCTL model checking – Summary

- Computation of set $\text{Sat}(\Phi)$ for DTMC D and PCTL formula Φ
 - recursive descent of parse tree
 - combination of graph algorithms, numerical computation
- Probabilistic operator P :
 - $X \Phi$: one matrix–vector multiplication, $O(|S|^2)$
 - $\Phi_1 U^{\leq k} \Phi_2$: k matrix–vector multiplications, $O(k|S|^2)$
 - $\Phi_1 U \Phi_2$: linear equation system, at most $|S|$ variables, $O(|S|^3)$
- Complexity:
 - linear in $|\Phi|$ and polynomial in $|S|$

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- **LTL model checking**
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Limitations of PCTL

- PCTL, although useful in practice, has limited expressivity
 - essentially: probability of reaching states in X , passing only through states in Y (and within k time-steps)
- More expressive logics can be used, for example:
 - LTL [Pnu77] – (non-probabilistic) linear-time temporal logic
 - PCTL* [ASB+95,BdA95] – which subsumes both PCTL and LTL
 - both allow path operators to be combined
 - (in PCTL, $P_{\sim p}[\dots]$ always contains a single temporal operator)
- Another direction: extend DTMCs with costs and rewards...

LTL – Linear temporal logic

- LTL syntax (path formulae only)

- $\psi ::= \text{true} \mid a \mid \psi \wedge \psi \mid \neg\psi \mid X\psi \mid \psi \cup \psi$
- where $a \in AP$ is an atomic proposition
- usual equivalences hold: $F\phi \equiv \text{true} \cup \phi$, $G\phi \equiv \neg(F\neg\phi)$

- LTL semantics (for a path ω)

- $\omega \models \text{true}$ always
- $\omega \models a$ $\Leftrightarrow a \in L(\omega(0))$
- $\omega \models \psi_1 \wedge \psi_2$ $\Leftrightarrow \omega \models \psi_1$ and $\omega \models \psi_2$
- $\omega \models \neg\psi$ $\Leftrightarrow \omega \not\models \psi$
- $\omega \models X\psi$ $\Leftrightarrow \omega[1\dots] \models \psi$
- $\omega \models \psi_1 \cup \psi_2$ $\Leftrightarrow \exists k \geq 0$ s.t. $\omega[k\dots] \models \psi_2 \wedge \forall i < k \omega[i\dots] \models \psi_1$

where $\omega(i)$ is i^{th} state of ω , and $\omega[i\dots]$ is suffix starting at $\omega(i)$

LTL examples

- $(F \text{ tmp_fail}_1) \wedge (F \text{ tmp_fail}_2)$
 - “both servers suffer temporary failures at some point”
- $GF \text{ ready}$
 - “the server always eventually returns to a ready-state”
- $FG \text{ error}$
 - “an irrecoverable error occurs”
- $G (\text{req} \rightarrow X \text{ ack})$
 - “requests are always immediately acknowledged”

LTL for DTMCs

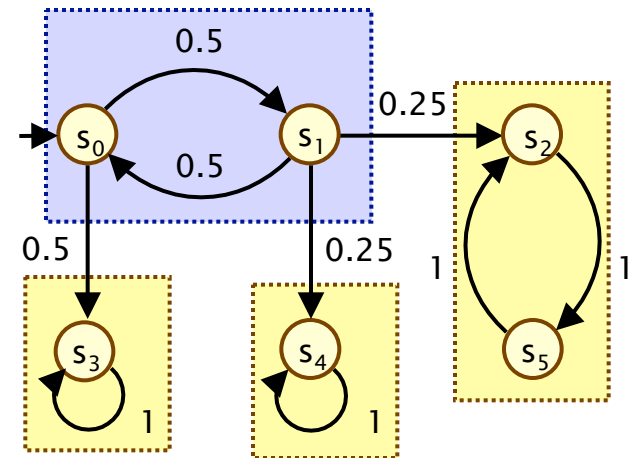
- Same idea as PCTL: probabilities of sets of path formulae
 - for a state s of a DTMC and an LTL formula ψ :
 - $\text{Prob}(s, \psi) = \Pr_s \{ \omega \in \text{Path}(s) \mid \omega \models \psi \}$
 - all such path sets are measurable [Var85]
- A (probabilistic) LTL specification often comprises an LTL (path) formula and a probability bound
 - e.g. $P_{\geq 1} [\text{GF ready}]$ – “with probability 1, the server always eventually returns to a ready-state”
 - e.g. $P_{<0.01} [\text{FG error}]$ – “with probability at most 0.01, an irrecoverable error occurs”
- PCTL* subsumes both LTL and PCTL
 - e.g. $P_{>0.5} [\text{GF crit}_1] \wedge P_{>0.5} [\text{GF crit}_2]$

Fundamental property of DTMCs

- Strongly connected component (SCC)
 - maximally strongly connected set of states
- Bottom strongly connected component (BSCC)
 - SCC T from which no state outside T is reachable from T

- Fundamental property of DTMCs:

- “with probability 1, a BSCC will be reached and all of its states visited infinitely often”



- Formally:

- $\Pr_s \{ \omega \in \text{Path}(s) \mid \exists i \geq 0, \exists \text{ BSCC } T \text{ such that}$
 $\forall j \geq i \omega(j) \in T \text{ and}$
 $\forall s' \in T \omega(k) = s' \text{ for infinitely many } k \} = 1$

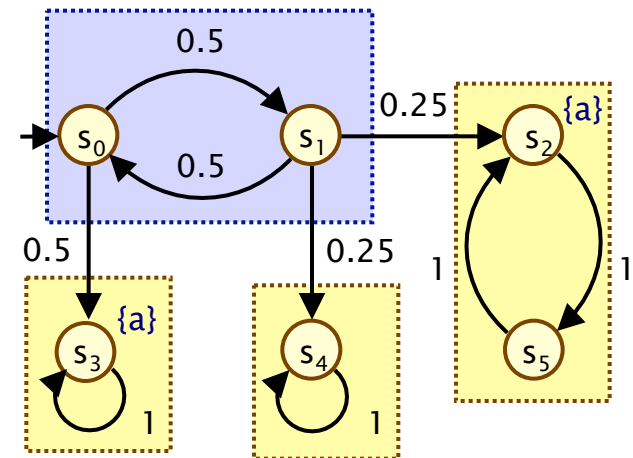
LTL model checking for DTMCs

- LTL model checking for DTMCs relies on:
 - computing probability of reaching a set of “accepting” BSCCs
 - e.g. for two simple LTL formulae: **GF a** (“always eventually a”), **FG a** (“eventually always a”) we have:

- $\text{Prob}(s, \text{GF } a) = \text{Prob}(s, F T_{\text{GF}a})$
 - where $T_{\text{GF}a}$ = union of all BSCCs containing some state satisfying a

- $\text{Prob}(s, \text{FG } a) = \text{Prob}(s, F T_{\text{FG}a})$
 - where $T_{\text{FG}a}$ = union of all BSCCs containing only a-states

- To extend this idea to arbitrary LTL formula, we use ω -automata...



Example:

$$\begin{aligned} \text{Prob}(s_0, \text{GF } a) &= \text{Prob}(s_0, F T_{\text{GF}a}) \\ &= \text{Prob}(s_0, F \{s_3, s_2, s_5\}) \\ &= 2/3 + 1/6 = 5/6 \end{aligned}$$

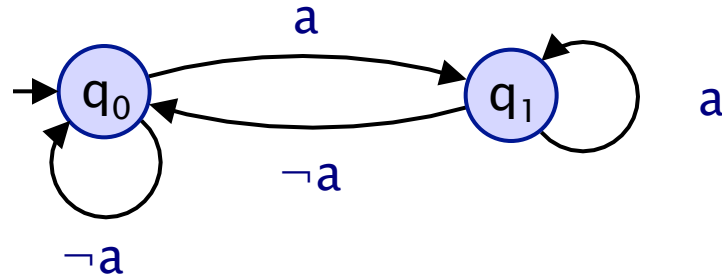
Deterministic Rabin automata

- ω -automata represent sets of **infinite** words
 - e.g. Buchi automata, Rabin automata, ...
 - for probabilistic model checking, need **deterministic** automata
 - so we use deterministic Rabin automata (DRAs)
- A deterministic Rabin automaton is a tuple $(Q, \Sigma, \delta, q_0, \text{Acc})$:
 - Q is a finite set of states, $q_0 \in Q$ is an initial state
 - Σ is an alphabet, $\delta : Q \times \Sigma \rightarrow Q$ is a transition function
 - $\text{Acc} = \{ (L_i, K_i) \}_{i=1..k} \subseteq 2^Q \times 2^Q$ is an acceptance condition
- A run of a word on a DRA is accepting iff:
 - for some pair (L_i, K_i) , the states in L_i are visited finitely often and (some of) the states in K_i are visited infinitely often
 - or in LTL: $\bigvee_{1 \leq i \leq k} (\text{FG } \neg L_i \wedge \text{GF } K_i)$

LTL & DRAs

- Example: DRA for **FG a**

- acceptance condition is
 $\text{Acc} = \{ (\{q_0\}, \{q_1\}) \}$



- Can convert any LTL formula ψ on atomic propositions AP

- into an equivalent DRA A_ψ over alphabet 2^{AP}
- i.e. $\omega \models \psi \Leftrightarrow \text{trace}(\omega) \in L(A_\psi)$ for any path ω
- can potentially incur a double exponential blow-up (but, in practice, this does not occur and ψ is small anyway)

- LTL model checking for DTMCs – the basic idea

- construct product of DTMC D and DRA A_ψ
- compute $\text{Prob}^D(s, \psi)$ on product DTMC $D \otimes A$

Product DTMC for a DRA

- The product DTMC $D \otimes A$ for:

- for DTMC $D = (S, s_{init}, P, L)$ and

- and (total) DRA $A = (Q, \Sigma, \delta, q_0, \{ (L_i, K_i) \}_{i=1..k})$

- is the DTMC $(S \times Q, (s_{init}, q_{init}), P', L')$ where:

$$q_{init} = \delta(q_0, L(s_{init}))$$

$$P'((s_1, q_1), (s_2, q_2)) = \begin{cases} P(s_1, s_2) & \text{if } q_2 = \delta(q_1, L(s_2)) \\ 0 & \text{otherwise} \end{cases}$$

$$l_i \in L'(s, q) \text{ if } q \in L_i \text{ and } k_i \in L'(s, q) \text{ if } q \in K_i$$

- Note:

- $D \otimes A$ can be seen as unfolding of D where q for each state (s, q) records state of automata A for path fragment so far

- since A is deterministic, $D \otimes A$ is a DTMC

- each path in D has a corresponding (unique) path in $D \otimes A$

- the probabilities of paths in D are preserved in $D \otimes A$

Product DTMC for a DRA

- For DTMC **D** and DRA **A**

$$\text{Prob}^D(s, A) = \text{Prob}^{D \otimes A}((s, q_s), \bigvee_{1 \leq i \leq k} (\text{FG } \neg l_i \wedge \text{GF } k_i))$$

– where $q_s = \delta(q_0, L(s))$

- Hence:

$$\text{Prob}^D(s, A) = \text{Prob}^{D \otimes A}((s, q_s), F T_{\text{Acc}})$$

– where T_{Acc} is the union of all **accepting BSCCs** in $D \otimes A$

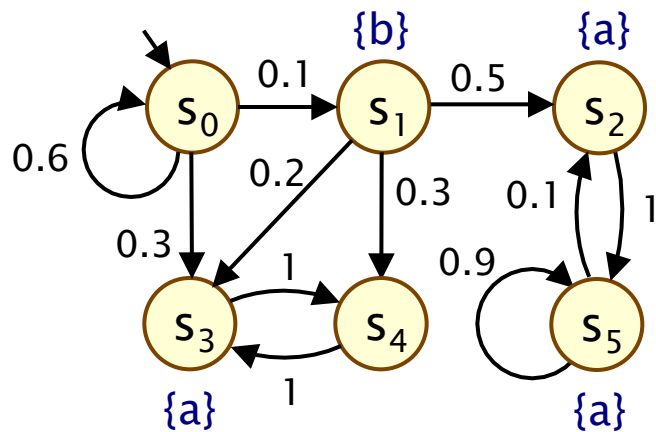
– an **accepting BSCC** T of $D \otimes A$ is such that, for some $1 \leq i \leq k$, no states in T satisfy l_i and some state in T satisfies k_i

- Reduces to computing BSCCs and reachability probabilities
 - so overall complexity for LTL is doubly exponential in $|\psi|$, polynomial in $|M|$; but can be reduced to singly exponential

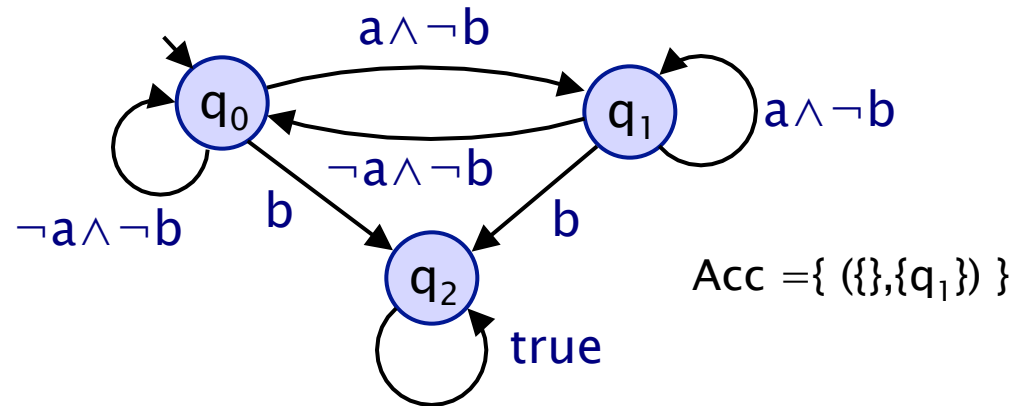
Example: LTL for DTMCs

- Compute $\text{Prob}(s_0, G\neg b \wedge GF a)$ for DTMC D :

DTMC D

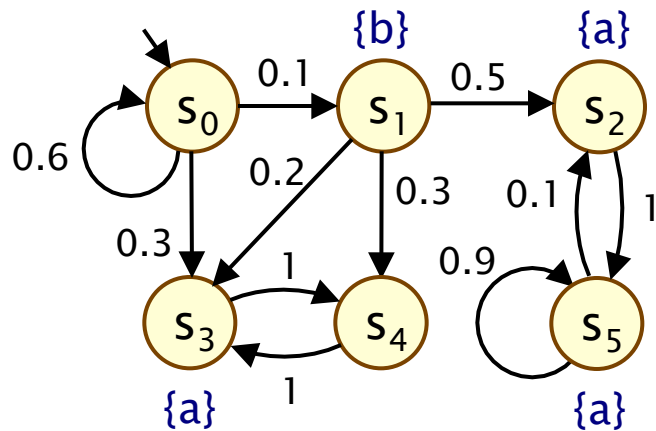


DRA A_ψ for $\psi = G\neg b \wedge GF a$

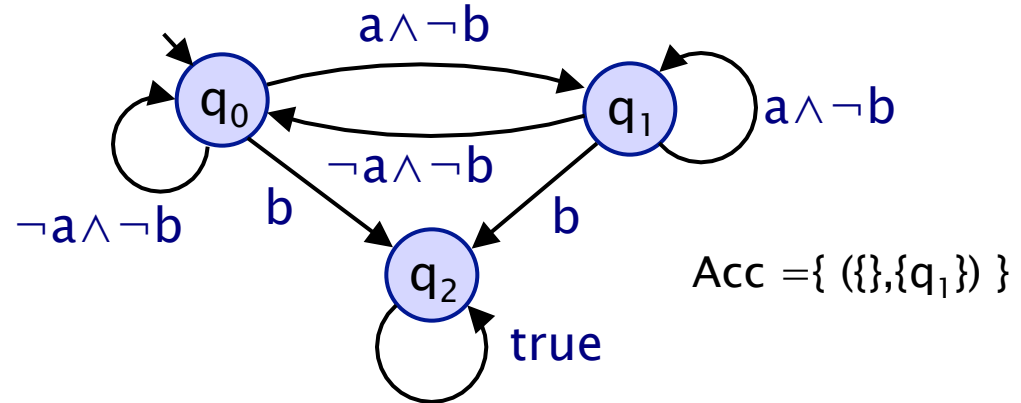


Example: LTL for DTMCs

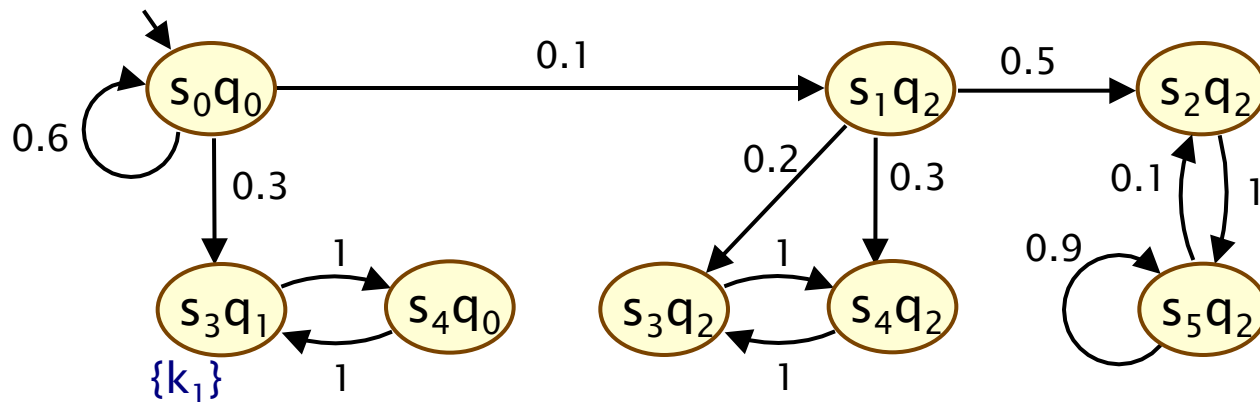
DTMC **D**



DRA A_ψ for $\psi = G\neg b \wedge GF a$

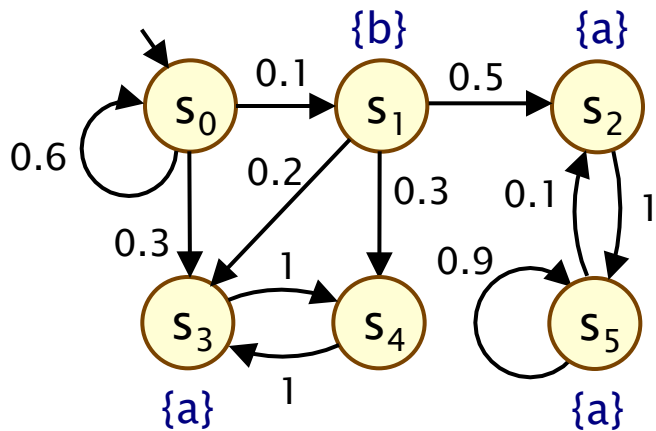


Product DTMC $D \otimes A_\psi$

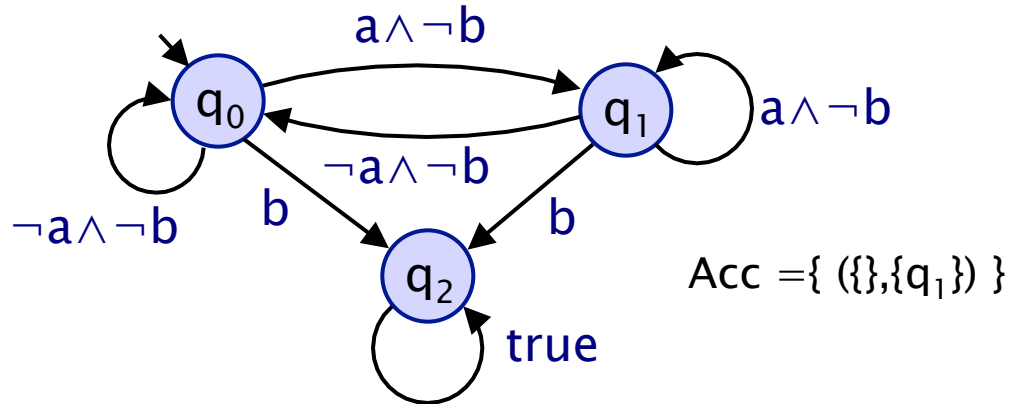


Example: LTL for DTMCs

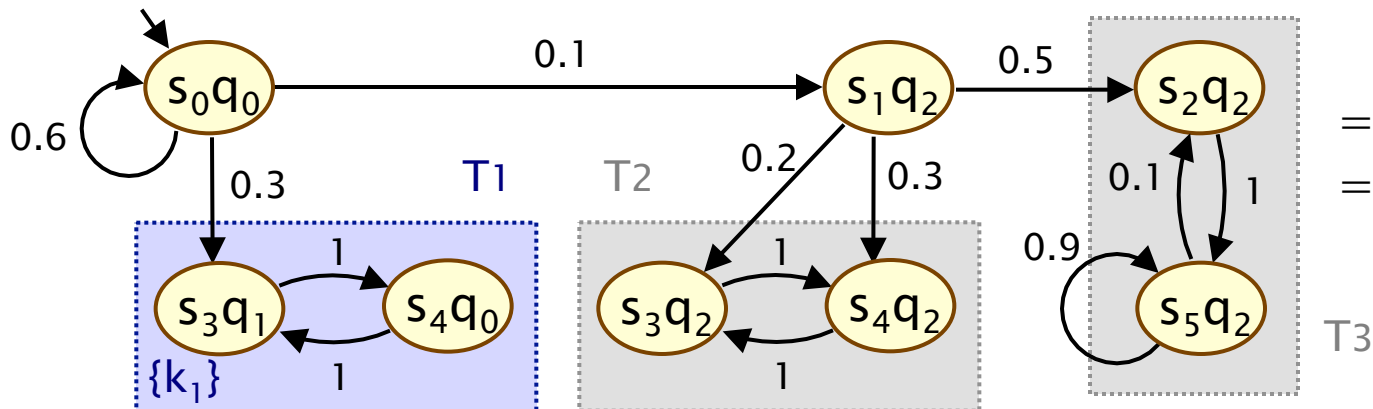
DTMC D



DRA A_ψ for $\psi = G\neg b \wedge GF a$



Product DTMC $D \otimes A_\psi$



$$\begin{aligned} \text{Prob}^D(s, \psi) &= \text{Prob}^{D \otimes A_\psi}(F T_1) \\ &= 3/4. \end{aligned}$$

Overview (Part 1)

- Discrete-time Markov chains (DTMCs)
- PCTL: A temporal logic for DTMCs
- PCTL model checking
- LTL model checking
- Costs and rewards

Costs and rewards

- We augment DTMCs with rewards (or, conversely, costs)
 - real-valued quantities assigned to states and/or transitions
 - these can have a wide range of possible interpretations
- Some examples:
 - elapsed time, power consumption, size of message queue, number of messages successfully delivered, net profit, ...
- Costs? or rewards?
 - mathematically, no distinction between rewards and costs
 - when interpreted, we assume that it is desirable to minimise costs and to maximise rewards
 - we will consistently use the terminology “rewards” regardless

Reward-based properties

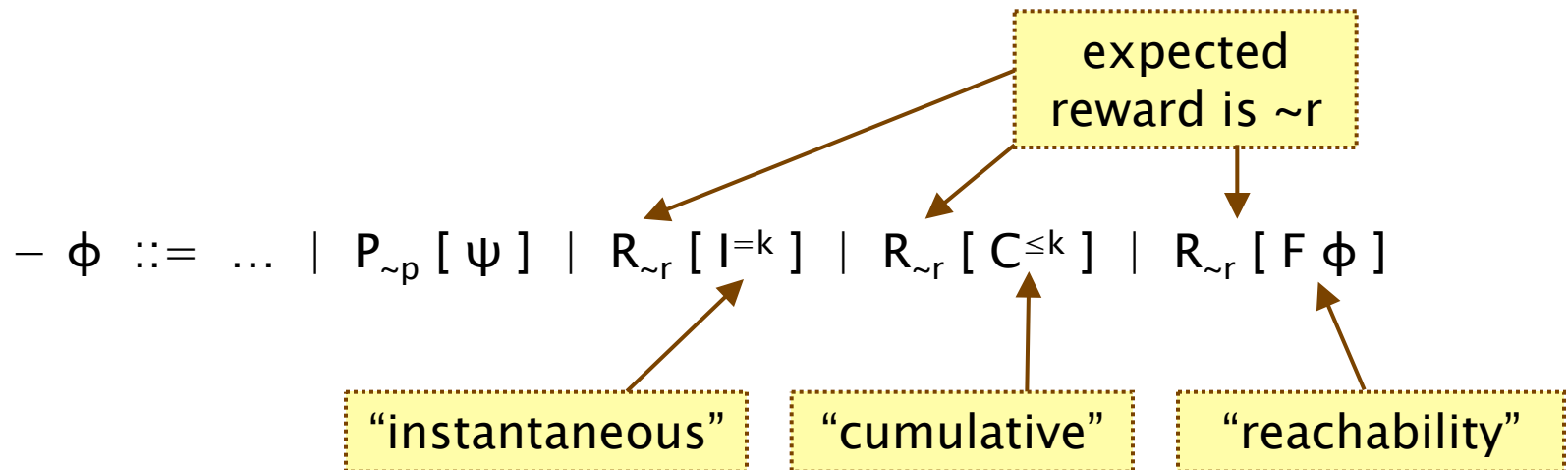
- Properties of DTMCs augmented with rewards
 - allow a wide range of quantitative measures of the system
 - basic notion: expected value of rewards
 - formal property specifications will be in an extension of PCTL
- More precisely, we use two distinct classes of property...
- **Instantaneous** properties
 - the expected value of the reward at some time point
- **Cumulative** properties
 - the expected cumulated reward over some period

DTMC reward structures

- For a DTMC (S, s_{init}, P, L) , a reward structure is a pair $(\underline{\rho}, \underline{\iota})$
 - $\underline{\rho} : S \rightarrow \mathbb{R}_{\geq 0}$ is the **state reward function** (vector)
 - $\underline{\iota} : S \times S \rightarrow \mathbb{R}_{\geq 0}$ is the **transition reward function** (matrix)
- Example (for use with instantaneous properties)
 - “size of message queue”: $\underline{\rho}$ maps each state to the number of jobs in the queue in that state, $\underline{\iota}$ is not used
- Examples (for use with cumulative properties)
 - “**time-steps**”: $\underline{\rho}$ returns 1 for all states and $\underline{\iota}$ is zero (equivalently, $\underline{\rho}$ is zero and $\underline{\iota}$ returns 1 for all transitions)
 - “**number of messages lost**”: $\underline{\rho}$ is zero and $\underline{\iota}$ maps transitions corresponding to a message loss to 1
 - “**power consumption**”: $\underline{\rho}$ is defined as the per-time-step energy consumption in each state and $\underline{\iota}$ as the energy cost of each transition

PCTL and rewards

- Extend PCTL to incorporate reward-based properties
 - add an R operator, which is similar to the existing P operator



– where $r \in \mathbb{R}_{\geq 0}$, $\sim \in \{<, >, \leq, \geq\}$, $k \in \mathbb{N}$

- $R_{\sim r}[\cdot]$ means “the **expected value** of \cdot satisfies $\sim r$ ”

Types of reward formulas

- **Instantaneous:** $R_{\sim r} [I^k]$
 - “the expected value of the state reward at time-step k is $\sim r$ ”
 - e.g. “the expected queue size after exactly 90 seconds”
- **Cumulative:** $R_{\sim r} [C^{\leq k}]$
 - “the expected reward cumulated up to time-step k is $\sim r$ ”
 - e.g. “the expected power consumption over one hour”
- **Reachability:** $R_{\sim r} [F \phi]$
 - “the expected reward cumulated before reaching a state satisfying ϕ is $\sim r$ ”
 - e.g. “the expected time for the algorithm to terminate”

Reward formula semantics

- Formal semantics of the three reward operators
 - based on random variables over (infinite) paths
- Recall:
 - $s \models P_{\sim p} [\psi] \Leftrightarrow \Pr_s \{ \omega \in \text{Path}(s) \mid \omega \models \psi \} \sim p$
- For a state s in the DTMC:
 - $s \models R_{\sim r} [I^k] \Leftrightarrow \text{Exp}(s, X_{I^k}) \sim r$
 - $s \models R_{\sim r} [C^{\leq k}] \Leftrightarrow \text{Exp}(s, X_{C^{\leq k}}) \sim r$
 - $s \models R_{\sim r} [F \Phi] \Leftrightarrow \text{Exp}(s, X_{F\Phi}) \sim r$

where: $\text{Exp}(s, X)$ denotes the **expectation** of the **random variable** $X : \text{Path}(s) \rightarrow \mathbb{R}_{\geq 0}$ with respect to the **probability measure** \Pr_s

Reward formula semantics

- Definition of random variables:
 - for an infinite path $\omega = s_0 s_1 s_2 \dots$

$$X_{I=k}(\omega) = \underline{\rho}(s_k)$$

$$X_{C \leq k}(\omega) = \begin{cases} 0 & \text{if } k = 0 \\ \sum_{i=0}^{k-1} \underline{\rho}(s_i) + \iota(s_i, s_{i+1}) & \text{otherwise} \end{cases}$$

$$X_{F\phi}(\omega) = \begin{cases} 0 & \text{if } s_0 \in \text{Sat}(\phi) \\ \infty & \text{if } s_i \notin \text{Sat}(\phi) \text{ for all } i \geq 0 \\ \sum_{i=0}^{k_\phi-1} \underline{\rho}(s_i) + \iota(s_i, s_{i+1}) & \text{otherwise} \end{cases}$$

- where $k_\phi = \min\{j \mid s_j \models \phi\}$

Model checking reward properties

- Instantaneous: $R_{\sim r} [I^k]$
- Cumulative: $R_{\sim r} [C^{\leq t}]$
 - variant of the method for computing bounded until probabilities
 - solution of **recursive equations**
- Reachability: $R_{\sim r} [F \phi]$
 - similar to computing until probabilities
 - precomputation phase (identify infinite reward states)
 - then reduces to solving a **system of linear equation**
- For more details, see e.g. [\[KNP07a\]](#)

Summary

- Probabilistic model checking
 - automated quantitative verification of stochastic systems
 - to model randomisation, failures, ...
- Discrete-time Markov chains (DTMCs)
 - state transition systems + discrete probabilistic choice
 - probability space over paths through a DTMC
- Property specifications
 - probabilistic extensions of temporal logic, e.g. PCTL, LTL
 - also: expected value of costs/rewards
- Model checking algorithms
 - combination of graph-based algorithms, numerical computation, automata constructions
- Tomorrow: Markov decision processes (MDPs)