

# Quantitative Automata Models and Model Checking

Joost-Pieter Katoen

RWTH Aachen University  
Software Modeling and Verification Group

SFM 2013 Summerschool on Dynamical Systems, Bertinoro, Italy

June 18, 2013



## Probabilities help

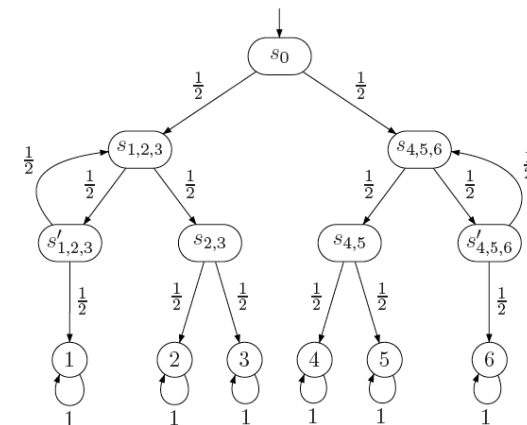
- ▶ When analysing system performance and dependability
  - ▶ to quantify arrivals, waiting times, time between failure, QoS, ...
- ▶ When modelling unreliable and unpredictable system behavior
  - ▶ to quantify message loss, processor failure
  - ▶ to quantify unpredictable delays, express soft deadlines, ...
- ▶ When building protocols for networked embedded systems
  - ▶ randomized algorithms
- ▶ When problems are undecidable deterministically
  - ▶ repeated reachability of lossy channel systems, ...

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

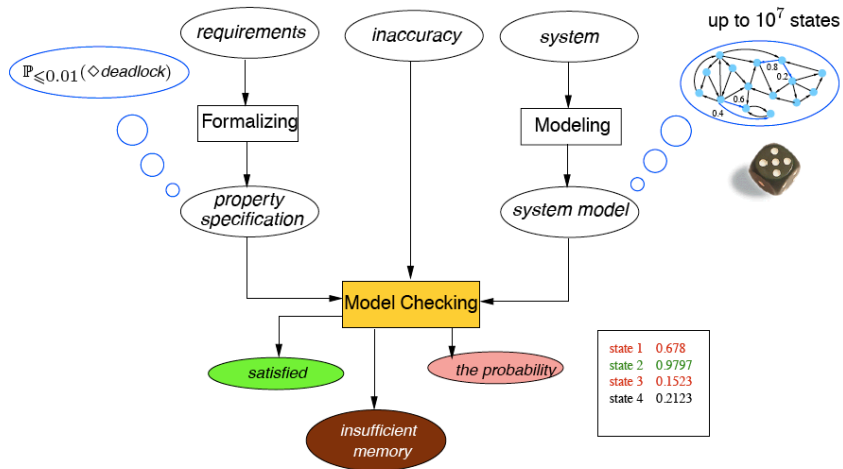
## Simulating a die by a fair coin

[Knuth & Yao]



Heads = “go left”; tails = “go right”. Does this DTMC model a six-sided die?

## What is probabilistic model checking?



## Properties

	Logic	Monitors
Discrete time	probabilistic CTL	deterministic automata (safety and LTL)
Continuous time	probabilistic timed CTL	deterministic timed automata

Core problem: computing (timed) reachability probabilities

## Probabilistic models

	Nondeterminism no	Nondeterminism yes
Discrete time	discrete-time Markov chain (DTMC)	Markov decision process (MDP)
Continuous time	CTMC	CTMDP

Some other models: probabilistic variants of (priced) timed automata

## Probability theory is simple, isn't it?

*In no other branch of mathematics  
is it so easy to make mistakes  
as in probability theory*

Henk Tijms, "Understanding Probability" (2004)



## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

## Memoryless property

### Theorem

1. For any random variable  $X$  with a geometric distribution:

$$Pr\{X = k + m \mid X > m\} = Pr\{X = k\} \quad \text{for any } m \in \mathbb{T}, k \geq 1$$

This is called the **memoryless** property, and  $X$  is a **memoryless r.v.**

2. **Any** discrete random variable which is memoryless is geometrically distributed.

## Geometric distribution

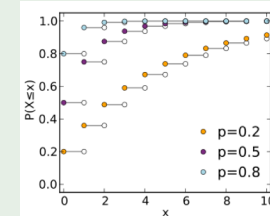
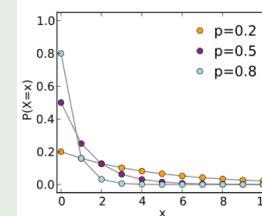
### Geometric distribution

Let  $X$  be a discrete random variable, natural  $k > 0$  and  $0 < p \leq 1$ . The mass function of a **geometric distribution** is given by:

$$Pr\{X = k\} = (1 - p)^{k-1} \cdot p$$

We have  $E[X] = \frac{1}{p}$  and  $Var[X] = \frac{1-p}{p^2}$  and cdf  $Pr\{X \leq k\} = 1 - (1-p)^k$ .

### Geometric distributions and their cdf's



## Markov property

The conditional probability distribution of future states of a Markov process only depends on the current state and not on its further history.

### Markov process

A discrete-time stochastic process  $\{X(t) \mid t \in \mathbb{T}\}$  over state space  $\{d_0, d_1, \dots\}$  is a **Markov process** if for any  $t_0 < t_1 < \dots < t_n < t_{n+1}$ :

$$\begin{aligned} Pr\{X(t_{n+1}) = d_{n+1} \mid X(t_0) = d_0, X(t_1) = d_1, \dots, X(t_n) = d_n\} \\ = \\ Pr\{X(t_{n+1}) = d_{n+1} \mid X(t_n) = d_n\} \end{aligned}$$

The distribution of  $X(t_{n+1})$ , given the values  $X(t_0)$  through  $X(t_n)$ , only depends on the current state  $X(t_n)$ .

## Invariance to time-shifts

### Time homogeneity

Markov process  $\{X(t) \mid t \in T\}$  is *time-homogeneous* iff for any  $t' < t$ :

$$\Pr\{X(t) = d \mid X(t') = d'\} = \Pr\{X(t - t') = d \mid X(0) = d'\}.$$

A time-homogeneous stochastic process is invariant to time shifts.

### Discrete-time Markov chain

A *discrete-time Markov chain* (DTMC) is a time-homogeneous Markov process with discrete parameter  $T$  and discrete state space  $S$ .

## Transition probability matrix

### Discrete-time Markov chain

A *discrete-time Markov chain* (DTMC) is a time-homogeneous Markov process with discrete parameter  $T$  and discrete state space  $S$ .

### Transition probability matrix

Let  $\mathbf{P}$  be a function with  $\mathbf{P}(s_i, s_j) = p(s_i, s_j)$ . For finite state space  $S$ , function  $\mathbf{P}$  is called the *transition probability matrix* of the DTMC with state space  $S$ .

### Properties

1.  $\mathbf{P}$  is a (right) *stochastic* matrix, i.e., it is a square matrix, all its elements are in  $[0, 1]$ , and each row sum equals one.
2.  $\mathbf{P}$  has an eigenvalue of one, and all its eigenvalues are at most one.
3. For all  $n \in \mathbb{N}$ ,  $\mathbf{P}^n$  is a stochastic matrix.

## Discrete-time Markov chain

### Discrete-time Markov chain

A *discrete-time Markov chain* (DTMC) is a time-homogeneous Markov process with discrete parameter  $T$  and discrete state space  $S$ .

### Transition probabilities

The *(one-step) transition probability* from  $s \in S$  to  $s' \in S$  at epoch  $n \in \mathbb{N}$  is given by:

$$p^{(n)}(s, s') = \Pr\{X_{n+1} = s' \mid X_n = s\} = \Pr\{X_1 = s' \mid X_0 = s\}$$

where the last equality is due to time-homogeneity.

Since  $p^{(n)}(\cdot) = p^{(k)}(\cdot)$ , the superscript  $(n)$  is omitted, and we write  $p(\cdot)$ .

## DTMCs — A transition system perspective

### Discrete-time Markov chain

A *DTMC*  $\mathcal{D}$  is a tuple  $(S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  with:

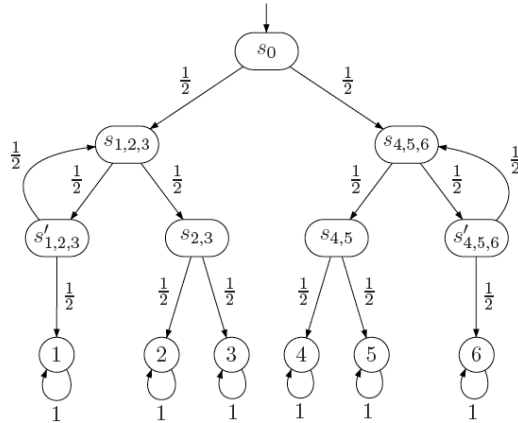
- ▶  $S$  is a countable nonempty set of *states*
- ▶  $\mathbf{P} : S \times S \rightarrow [0, 1]$ , *transition probability function* s.t.  $\sum_{s'} \mathbf{P}(s, s') = 1$
- ▶  $\iota_{\text{init}} : S \rightarrow [0, 1]$ , the *initial distribution* with  $\sum_{s \in S} \iota_{\text{init}}(s) = 1$
- ▶  $AP$  is a set of *atomic propositions*.
- ▶  $L : S \rightarrow 2^{AP}$ , the *labeling function*, assigning to state  $s$ , the set  $L(s)$  of atomic propositions that are valid in  $s$ .

### Initial states

- ▶  $\iota_{\text{init}}(s)$  is the probability that DTMC  $\mathcal{D}$  starts in state  $s$
- ▶ the set  $\{s \in S \mid \iota_{\text{init}}(s) > 0\}$  are the possible *initial states*.

## Simulating a die by a fair coin

[Knuth &amp; Yao]



Heads = “go left”; tails = “go right”. Does this DTMC model a six-sided die?

## Determining $n$ -step transition probabilities

### $n$ -step transition probabilities

The probability to move from  $s$  to  $s'$  in  $n \in \mathbb{N}$  steps is inductively defined:

$$p_{s,s'}(0) = 1 \quad \text{if } s = s', \quad \text{and } 0 \text{ otherwise,}$$

$p_{s,s'}(1) = \mathbf{P}(s, s')$ , and for  $n > 1$  by the Chapman-Kolmogorov equation:

$$p_{s,s'}(n) = \sum_{s''} p_{s,s''}(l) \cdot p_{s'',s'}(n-l) \quad \text{for some } 0 < l < n$$

For  $l = 1$  and  $n > 0$  we obtain:  $p_{s,s'}(n) = \sum_{s''} p_{s,s''}(1) \cdot p_{s'',s'}(n-1)$

$\mathbf{P}^{(n)} = \mathbf{P}^{(1)} \cdot \mathbf{P}^{(n-1)} = \mathbf{P} \cdot \mathbf{P}^{(n-1)}$  is the  $n$ -step transition probability matrix

Repeating this scheme:  $\mathbf{P}^{(n)} = \mathbf{P} \cdot \mathbf{P}^{(n-1)} = \dots = \mathbf{P}^{n-1} \cdot \mathbf{P}^{(1)} = \mathbf{P}^n$ .

## State residence time distribution

Let  $T_s$  be the number of epochs of DTMC  $\mathcal{D}$  to **stay** in state  $s$ :

$$Pr\{T_s = 1\} = 1 - \mathbf{P}(s, s)$$

$$Pr\{T_s = 2\} = \mathbf{P}(s, s) \cdot (1 - \mathbf{P}(s, s))$$

.....

$$Pr\{T_s = n\} = \mathbf{P}(s, s)^{n-1} \cdot (1 - \mathbf{P}(s, s))$$

So, the state residence times in a DTMC obey a *geometric* distribution.

The expected number of time steps to stay in state  $s$  equals  $E[T_s] = \frac{1}{1 - \mathbf{P}(s, s)}$ .

The variance of the residence time distribution is  $Var[T_s] = \frac{\mathbf{P}(s, s)}{(1 - \mathbf{P}(s, s))^2}$ .

A geometric distribution is the **only** discrete probability distribution that is memoryless.

## Transient probability distribution

### Transient distribution

$\mathbf{P}^n(s, t)$  equals the probability of being in state  $t$  after  $n$  steps given that the computation starts in  $s$ .

The probability of DTMC  $\mathcal{D}$  being in state  $t$  after exactly  $n$  transitions is:

$$\Theta_n^{\mathcal{D}}(t) = \sum_{s \in S} \iota_{\text{init}}(s) \cdot \mathbf{P}^n(s, t)$$

$\Theta_n^{\mathcal{D}}(t)$  is called the *transient state probability* at epoch  $n$  for state  $t$ . The function  $\Theta_n^{\mathcal{D}}$  is the *transient state distribution* at epoch  $n$  of DTMC  $\mathcal{D}$ .

When considering  $\Theta_n^{\mathcal{D}}$  as vector  $(\Theta_n^{\mathcal{D}})_{t \in S}$  we have:

$$\Theta_n^{\mathcal{D}} = \iota_{\text{init}} \cdot \underbrace{\mathbf{P} \cdot \mathbf{P} \cdot \dots \cdot \mathbf{P}}_{n \text{ times}} = \iota_{\text{init}} \cdot \mathbf{P}^n.$$

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 **Reachability probabilities**
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

## Aim of this lecture

How to determine **reachability** probabilities?

### Three major steps

1. What are reachability probabilities? I mean, **precisely**.  
This requires a bit of **measure theory**. Sorry for that.
2. Reachability probabilities = unique solution of linear equation system.
3. ... and they are transient probabilities in a slightly modified DTMC.

## Summary

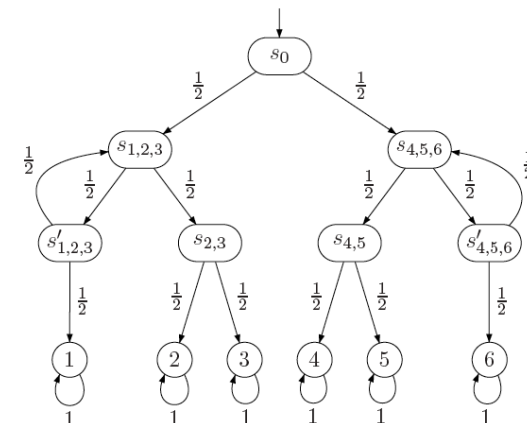
### What are Markov chains?

- ▶ A **discrete-time Markov chain** (DTMC) is a time-homogeneous Markov process with discrete parameter  $T$  and discrete state space  $S$ .
- ▶ State residence times are geometrically distributed.
- ▶ Alternative: a DTMC  $\mathcal{D}$  is a tuple  $(S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  with:
  - ▶ state space  $S$
  - ▶ transition probability function  $\mathbf{P}$
  - ▶ initial distribution  $\iota_{\text{init}}$

### What are transient probabilities?

- ▶  $\Theta_n^{\mathcal{D}}(s)$  is the probability to be in state  $s$  after  $n$  steps.
- ▶ These **transient probabilities** satisfy:  $\Theta_n^{\mathcal{D}} = \iota_{\text{init}} \cdot \mathbf{P}^n$ .

## Recall Knuth's die



Heads = "go left"; tails = "go right". Does this DTMC model a six-sided die?

## Paths

### State graph

The *state graph* of DTMC  $\mathcal{D}$  is a digraph  $G = (V, E)$  with  $V$  the states of  $\mathcal{D}$ , and  $(s, s') \in E$  iff  $\mathbf{P}(s, s') > 0$ .

Let  $Pre(s)$  be the *predecessors* of  $s$ ,  $Pre^*(s)$  its reflexive and transitive closure.

### Paths

*Paths* in  $\mathcal{D}$  are infinite paths in its state graph.

$Paths(\mathcal{D})$  denotes the set of paths in  $\mathcal{D}$ , and  $Paths^*(\mathcal{D})$  its finite prefixes.

## More events of interest

### Repeated reachability

Repeatedly visit a state in  $G$ ; formally:

$$\Box\Diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \forall i \in \mathbb{N}. \exists j \geq i. \pi[j] \in G \}$$

### Persistence

Eventually reach in a state in  $G$  and always stay there; formally:

$$\Diamond\Box G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \forall j \geq i. \pi[j] \in G \}$$

## Some events of interest

Let DTMC  $\mathcal{D}$  with (possibly infinite) state space  $S$ .

### (Simple) reachability

Eventually reach a state in  $G \subseteq S$ . Formally:

$$\Diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \}$$

Invariance, i.e., always stay in state in  $G$ :

$$\Box G = \{ \pi \in Paths(\mathcal{D}) \mid \forall i \in \mathbb{N}. \pi[i] \in G \} = \overline{\overline{\Diamond \bar{G}}}.$$

### Constrained reachability

Or “reach-avoid” properties where states in  $F \subseteq S$  are forbidden:

$$\bar{F} U G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \wedge \forall j < i. \pi[j] \notin F \}$$

## What's the probability of infinite paths?



## Paths and probabilities

To reason quantitatively about the behavior of a DTMC, we need to define a **probability space** over its paths.

### Intuition

For a given state  $s$  in DTMC  $\mathcal{D}$ :

- ▶ Outcomes := set of all infinite paths starting in  $s$ .
- ▶ Events := subsets of these outcomes.
- ▶ These events are defined using **cylinder sets**.
- ▶ Cylinder set of a finite path := set of all its infinite continuations.

## Probability measure on DTMCs

### Cylinder set

The cylinder set of finite path  $\hat{\pi} = s_0 s_1 \dots s_n \in Paths^*(\mathcal{D})$  is defined by:

$$Cyl(\hat{\pi}) = \{ \pi \in Paths(\mathcal{D}) \mid \hat{\pi} \text{ is a prefix of } \pi \}$$

### Probability measure

$Pr$  is the unique **probability measure** defined by:

$$Pr(Cyl(s_0 \dots s_n)) = \iota_{\text{init}}(s_0) \cdot \mathbf{P}(s_0 s_1 \dots s_n)$$

where  $\mathbf{P}(s_0 s_1 \dots s_n) = \prod_{0 \leq i < n} \mathbf{P}(s_i, s_{i+1})$  for  $n > 0$  and  $\mathbf{P}(s_0) = \iota_{\text{init}}(s_0)$ .

## Probability measure on DTMCs

### Cylinder set

The **cylinder set** of finite path  $\hat{\pi} = s_0 s_1 \dots s_n \in Paths^*(\mathcal{D})$  is defined by:

$$Cyl(\hat{\pi}) = \{ \pi \in Paths(\mathcal{D}) \mid \hat{\pi} \text{ is a prefix of } \pi \}$$

The cylinder set spanned by finite path  $\hat{\pi}$  thus consists of all infinite paths that have prefix  $\hat{\pi}$ .

### Probability space of a DTMC

The set of events of the probability space DTMC  $\mathcal{D}$  contains all cylinder sets  $Cyl(\hat{\pi})$  where  $\hat{\pi}$  ranges over all finite paths in  $\mathcal{D}$ .

## Measurability

### Measurability theorem

Events  $\Diamond G$ ,  $\Box G$ ,  $\overline{F} \cup G$ ,  $\Box \Diamond G$  and  $\Diamond \Box G$  are **measurable** on any DTMC.

### Proof:

To show this, every event has to be expressed as allowed operations (complement and/or countable unions) of the events — our cylinder sets! — of a DTMC.

Note that  $\Box G = \overline{\Diamond \overline{G}}$  and  $\Diamond \Box G = \overline{\Box \Diamond \overline{G}}$ .

It remains to prove the measurability for the remaining three cases.



## Proof for $\Diamond G$

Which event does  $\Diamond G$  exactly mean?

the union of all cylinders  $Cyl(s_0 \dots s_n)$  where

$s_0 \dots s_n$  is a finite path in  $\mathcal{D}$  with  $s_0, \dots, s_{n-1} \notin G$  and  $s_n \in G$ , i.e.,

$$\Diamond G = \bigcup_{s_0 \dots s_n \in Paths^*(\mathcal{D}) \cap (S \setminus G)^* G} Cyl(s_0 \dots s_n)$$

Thus  $\Diamond G$  is measurable.

As all cylinder sets are pairwise disjoint, its probability is defined by:

$$\begin{aligned} Pr(\Diamond G) &= \sum_{s_0 \dots s_n \in Paths^*(\mathcal{D}) \cap (S \setminus G)^* G} Pr(Cyl(s_0 \dots s_n)) \\ &= \sum_{s_0 \dots s_n \in Paths^*(\mathcal{D}) \cap (S \setminus G)^* G} l_{init}(s_0) \cdot P(s_0 \dots s_n) \end{aligned}$$

A similar proof strategy applies to the case  $\overline{F} U G$ .

## Reachability probabilities in finite DTMCs

### Problem statement

Let  $\mathcal{D}$  be a DTMC with finite state space  $S$ ,  $s \in S$  and  $G \subseteq S$ .

Aim: determine  $Pr(s \models \Diamond G) = Pr_s(\Diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \Diamond G\}$

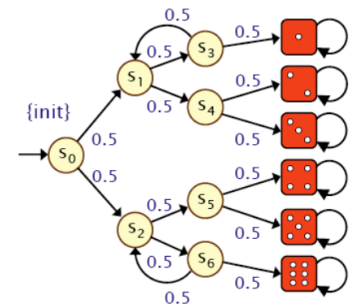
where  $Pr_s$  is the probability measure in  $\mathcal{D}$  with single initial state  $s$ .

### Characterisation of reachability probabilities

- Let variable  $x_s = Pr(s \models \Diamond G)$  for any state  $s$ 
  - if  $G$  is not reachable from  $s$ , then  $x_s = 0$
  - if  $s \in G$  then  $x_s = 1$
- For any state  $s \in Pre^*(G) \setminus G$ :

$$x_s = \underbrace{\sum_{t \in S \setminus G} P(s, t) \cdot x_t}_{\text{reach } G \text{ via } t \in S \setminus G} + \underbrace{\sum_{u \in G} P(s, u)}_{\text{reach } G \text{ in one step}}$$

## Reachability probabilities: Knuth's die



- Consider the event  $\Diamond 4$
- Using the previous theorem we obtain:

$$Pr(\Diamond 4) = \sum_{s_0 \dots s_n \in (S \setminus 4)^* 4} P(s_0 \dots s_n)$$

- This yields:  
 $P(s_0 s_2 s_5 4) + P(s_0 s_2 s_6 s_2 s_5 4) + \dots$

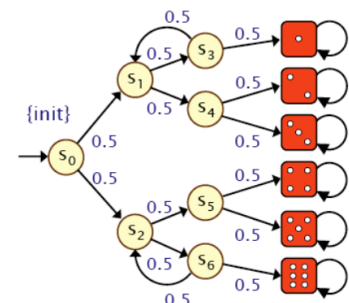
- Or:  $\sum_{k=0}^{\infty} P(s_0 s_2 (s_6 s_2)^k s_5 4)$

- Or:  $\frac{1}{8} \cdot \sum_{k=0}^{\infty} \left(\frac{1}{4}\right)^k$

- Geometric series:  $\frac{1}{8} \cdot \frac{1}{1 - \frac{1}{4}} = \frac{1}{8} \cdot \frac{4}{3} = \frac{1}{6}$

There is however an **simpler** way to obtain reachability probabilities!

## Reachability probabilities: Knuth's die



- Consider the event  $\Diamond 4$
- Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

$$x_{s_0} = \frac{1}{2}x_{s_1} + \frac{1}{2}x_{s_2}$$

$$x_{s_2} = \frac{1}{2}x_{s_5} + \frac{1}{2}x_{s_6}$$

$$x_{s_5} = \frac{1}{2}x_5 + \frac{1}{2}x_4$$

$$x_{s_6} = \frac{1}{2}x_{s_2} + \frac{1}{2}x_6$$

- Gaussian elimination yields:

$$x_{s_5} = \frac{1}{2}, x_{s_2} = \frac{1}{3}, x_{s_6} = \frac{1}{6}, \text{ and } \boxed{x_{s_0} = \frac{1}{6}}$$

## Linear equation system

### Reachability probabilities as linear equation system

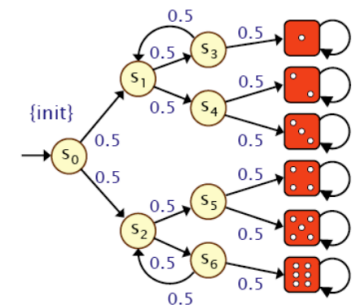
- ▶ Let  $S_? = Pre^*(G) \setminus G$ , the states that can reach  $G$  by  $> 0$  steps
- ▶  $\mathbf{A} = (\mathbf{P}(s, t))_{s, t \in S_?}$ , the transition probabilities in  $S_?$
- ▶  $\mathbf{b} = (b_s)_{s \in S_?}$ , the probs to reach  $G$  in 1 step, i.e.,  $b_s = \sum_{u \in G} \mathbf{P}(s, u)$

Then:  $\mathbf{x} = (x_s)_{s \in S_?}$  with  $x_s = Pr(s \models \Diamond G)$  is the **unique** solution of:

$$\mathbf{x} = \mathbf{A} \cdot \mathbf{x} + \mathbf{b} \quad \text{or} \quad (\mathbf{I} - \mathbf{A}) \cdot \mathbf{x} = \mathbf{b}$$

where  $\mathbf{I}$  is the identity matrix of cardinality  $|S_?| \times |S_?|$ .

## Reachability probabilities: Knuth's die



- ▶ Consider the event  $\Diamond 4$

$$S_? = \{s_0, s_2, s_5, s_6\}$$

$$\begin{pmatrix} 1 & -\frac{1}{2} & 0 & 0 \\ 0 & 1 & -\frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & 1 & 0 \\ 0 & -\frac{1}{2} & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_{s_0} \\ x_{s_2} \\ x_{s_5} \\ x_{s_6} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \frac{1}{2} \\ 0 \end{pmatrix}$$

- ▶ Gaussian elimination yields:

$$x_{s_5} = \frac{1}{2}, x_{s_2} = \frac{1}{3}, x_{s_6} = \frac{1}{6}, \text{ and } \boxed{x_{s_0} = \frac{1}{6}}$$

## Remark

### Iterative algorithms to compute $\mathbf{x}$

There are various algorithms to compute  $\mathbf{x} = \lim_{n \rightarrow \infty} \mathbf{x}^{(n)}$  where:

$$\mathbf{x}^{(0)} = \mathbf{0} \quad \text{and} \quad \mathbf{x}^{(i+1)} = \mathbf{A} \cdot \mathbf{x}^{(i)} + \mathbf{b} \quad \text{for } 0 \leq i.$$

Then:

1.  $\mathbf{x}^{(n)}(s) = Pr(s \models \Diamond^{\leq n} G)$  for  $s \in S_?$
2.  $\mathbf{x}^{(0)} \leq \mathbf{x}^{(1)} \leq \mathbf{x}^{(2)} \leq \dots \leq \mathbf{x}$  and  $\mathbf{x} = \lim_{n \rightarrow \infty} \mathbf{x}^{(n)}$

The **Power method** computes vectors  $\mathbf{x}^{(0)}, \mathbf{x}^{(1)}, \mathbf{x}^{(2)}, \dots$  and aborts if:

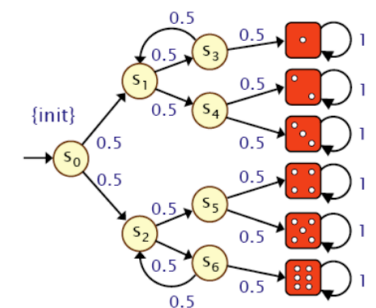
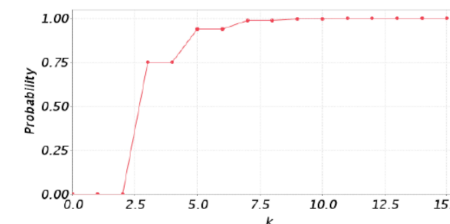
$$\max_{s \in S_?} |x_s^{(n+1)} - x_s^{(n)}| < \varepsilon \quad \text{for some small tolerance } \varepsilon$$

This technique guarantees **convergence**.

Alternatives: e.g., Jacobi or Gauss-Seidel, successive overrelaxation (SOR).

## Example: Knuth's die

- ▶ Let  $G = \{1, 2, 3, 4, 5, 6\}$
- ▶ Then  $Pr(s_0 \models \Diamond G) = 1$
- ▶ And  $Pr(s_0 \models \Diamond^{\leq k} G)$  for  $k \in \mathbb{N}$  is given by:



## Reachability probability = transient probabilities

### Aim

Compute  $Pr(\Diamond^{\leq n} G)$  in DTMC  $\mathcal{D}$ . Observe that once a path  $\pi$  reaches  $G$ , then the remaining behaviour along  $\pi$  is not important. This suggests to make all states in  $G$  absorbing.

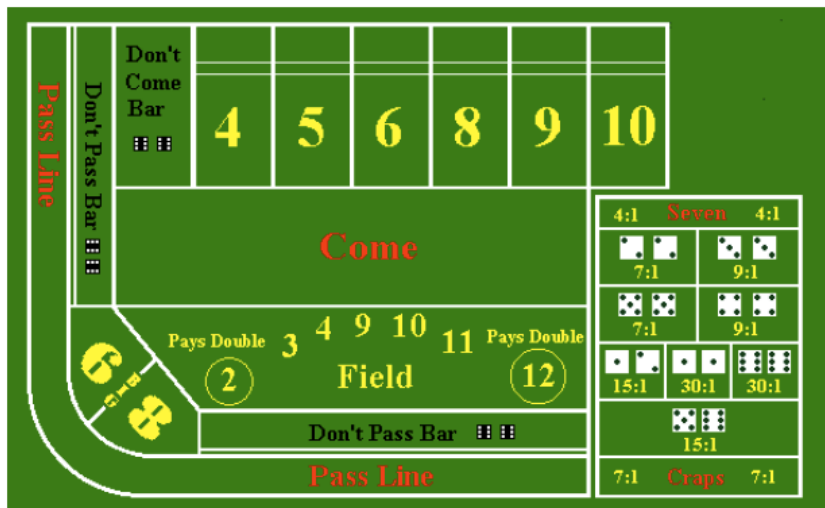
Let DTMC  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  and  $G \subseteq S$ . The DTMC  $\mathcal{D}[G] = (S, \mathbf{P}_G, \iota_{\text{init}}, AP, L)$  with  $\mathbf{P}_G(s, t) = \mathbf{P}(s, t)$  if  $s \notin G$  and  $\mathbf{P}_G(s, s) = 1$  if  $s \in G$ .

All outgoing transitions of  $s \in G$  are replaced by a single self-loop at  $s$ .

### Lemma

$$\underbrace{Pr(\Diamond^{\leq n} G)}_{\text{reachability in } \mathcal{D}} = \underbrace{Pr(\Diamond^{=n} G)}_{\text{reachability in } \mathcal{D}[G]} = \underbrace{\iota_{\text{init}} \cdot \mathbf{P}_G^n}_{\text{in } \mathcal{D}[G]} = \Theta_n^{\mathcal{D}[G]}$$

## Spare time tonight? Play Craps!



## Constrained reachability = transient probabilities

### Aim

Compute  $Pr(\overline{F} U^{\leq n} G)$  in DTMC  $\mathcal{D}$ . Observe (as before) that once a path  $\pi$  reaches  $G$  via  $\overline{F}$ , then the remaining behaviour along  $\pi$  is not important. Now also observe that once  $s \in F \setminus G$  is reached, then the remaining behaviour along  $\pi$  is not important. This suggests to make all states in  $G$  and  $F \setminus G$  absorbing.

### Lemma

$$\underbrace{Pr(\overline{F} U^{\leq n} G)}_{\text{reachability in } \mathcal{D}} = \underbrace{Pr(\Diamond^{=n} G)}_{\text{reachability in } \mathcal{D}[F \cup G]} = \underbrace{\iota_{\text{init}} \cdot \mathbf{P}_{F \cup G}^n}_{\text{in } \mathcal{D}[F \cup G]} = \Theta_n^{\mathcal{D}[F \cup G]}$$

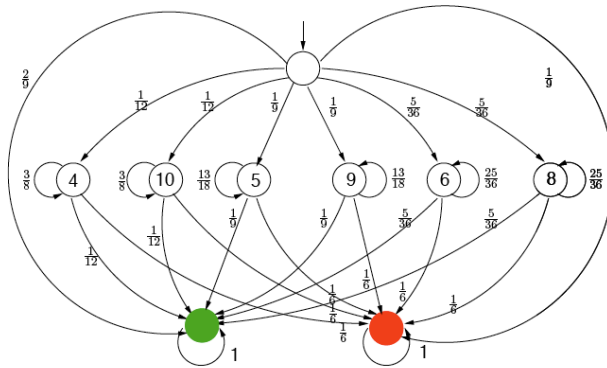
## Craps

- ▶ Roll two dice and bet
- ▶ Come-out roll ("pass line" wager):
  - ▶ outcome 7 or 11: win
  - ▶ outcome 2, 3, or 12: lose ("craps")
  - ▶ any other outcome: roll again (outcome is "point")
- ▶ Repeat until 7 or the "point" is thrown:
  - ▶ outcome 7: lose ("seven-out")
  - ▶ outcome the point: win
  - ▶ any other outcome: roll again



## A DTMC model of Craps

- ▶ Come-out roll:
  - ▶ 7 or 11: win
  - ▶ 2, 3, or 12: lose
  - ▶ else: roll again
- ▶ Next roll(s):
  - ▶ 7: lose
  - ▶ point: win
  - ▶ else: roll again



What is the probability to win the Craps game?

## Summary of previous lecture

How to determine **reachability** probabilities?

1. Probabilities of sets of infinite paths defined using **cylinders**.
2. Events  $\Diamond G$ ,  $\Box \Diamond G$  and  $\bar{F} U G$  are **measurable**.
3. Reachability probabilities = unique solution of **linear equation system**.
4. ... and they are **transient probabilities** in a slightly modified DTMC.

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

## Qualitative properties

### Quantitative properties

Comparing the probability of an event such as  $\Box G$ ,  $\Diamond \Box G$  and  $\Box \Diamond G$  with a threshold  $\sim p$  with  $p \in (0, 1)$  and  $\sim$  a binary comparison operator ( $=, <, \leq, \geq, >$ ) yields a **quantitative property**.

### Example quantitative properties

$$Pr(s \models \Diamond \Box G) > \frac{1}{2} \quad \text{or} \quad Pr(s \models \Diamond^{\leq n} G) \leq \frac{\pi}{5}$$

### Qualitative properties

Comparing the probability of an event such as  $\Box G$ ,  $\Diamond \Box G$  and  $\Box \Diamond G$  with a threshold  $> 0$  or  $= 1$  yields a **qualitative property**. Any event  $E$  with  $Pr(E) = 1$  is called **almost surely**.

### Example qualitative properties

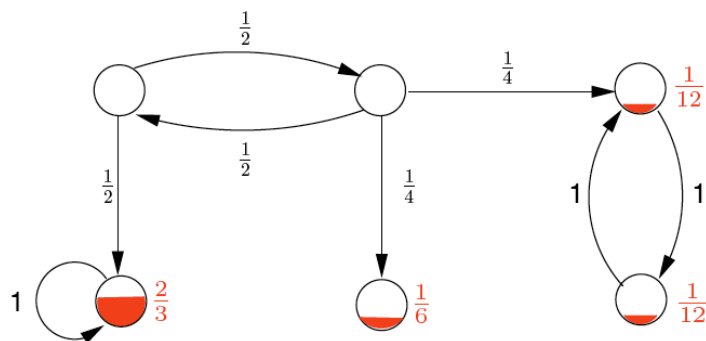
$$Pr(s \models \Diamond \Box G) > 0 \quad \text{or} \quad Pr(s \models \Diamond^{\leq n} G) = 1$$

## Verifying qualitative properties

### Remark

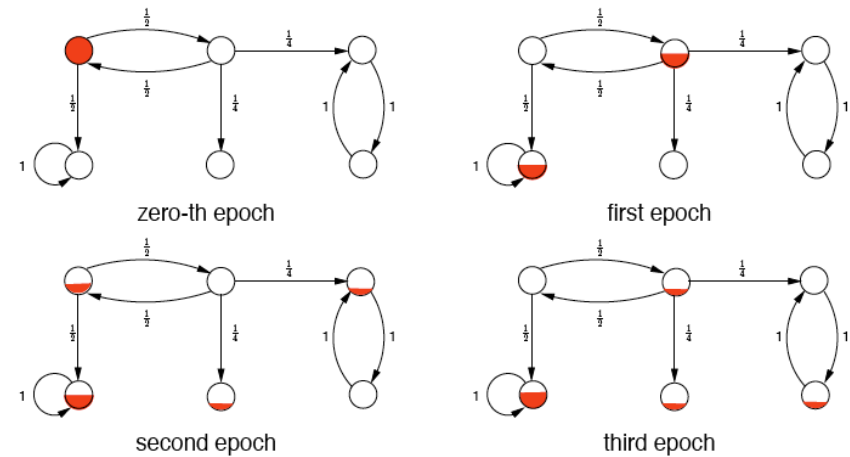
In the following we will concentrate on **almost sure** events, i.e., events  $E$  with  $Pr(E) = 1$ . This suffices, as  $Pr(E) > 0$  if and only if not  $Pr(\bar{E}) = 1$ .

## On the long run



The probability mass on the long run is only left in **bottom** SCCs.

## Where do we end up in the end?



Which states have a probability  $> 0$  when repeating this on the long run?

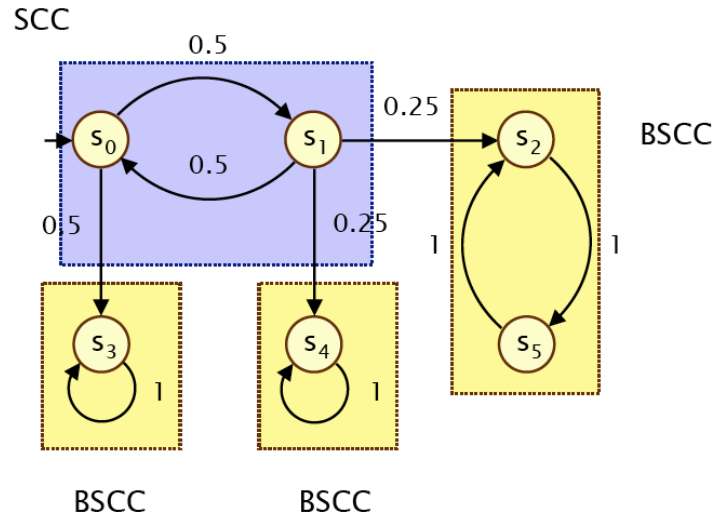
## What is a BSCC?

Let  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  be a (possibly infinite) DTMC.

### Strongly connected component

- ▶  $T \subseteq S$  is **strongly connected** if for any  $s, t \in T$ , states  $s$  and  $t \in T$  are mutually reachable via edges in  $T$ .
- ▶  $T$  is a **strongly connected component** (SCC) of  $\mathcal{D}$  if it is strongly connected and no proper superset of  $T$  is strongly connected.
- ▶ SCC  $T$  is a **bottom SCC** (BSCC) if no state outside  $T$  is reachable from  $T$ , i.e., for any state  $s \in T$ ,  $\mathbf{P}(s, T) = \sum_{t \in T} \mathbf{P}(s, t) = 1$ .

## Example



## Almost sure reachability

Recall: an absorbing state in a DTMC is a state with a self-loop with probability one.

### Almost sure reachability theorem

For finite DTMC with state space  $S$ ,  $s \in S$  and  $G \subseteq S$  a set of absorbing states:

$$Pr(s \models \Diamond G) = 1 \quad \text{iff} \quad s \in S \setminus Pre^*(S \setminus Pre^*(G)).$$

Note:  $S \setminus Pre^*(S \setminus Pre^*(G))$  are states that cannot reach states from which  $G$  cannot be reached.

### Proof:

Show that both sides of the equivalence are equivalent to  $Post^*(t) \cap G \neq \emptyset$  for each state  $t \in Post^*(s)$ . Rather straightforward.

## Long-run theorem

### Long-run theorem

For each state  $s$  of a finite Markov chain  $\mathcal{D}$ :

$$Pr_s\{\pi \in Paths(s) \mid \inf(\pi) \text{ is a BSCC of } \mathcal{D}\} = 1.$$

where  $\inf(\pi)$  is the set of states that are visited infinitely often along  $\pi$ .

### Intuition

Almost surely any finite DTMC eventually reaches a BSCC and visits all its states infinitely often.

### Remark

For any state  $s$  in (possibly infinite) DTMC  $\mathcal{D}$ :

$$\{\pi \in Paths(s) \mid \inf(\pi) \text{ is a BSCC of } \mathcal{D}\} \text{ is measurable.}$$

## Computing almost sure reachability properties

### Aim:

For finite DTMC  $\mathcal{D}$  and  $G \subseteq S$ , determine  $\{s \in S \mid Pr(s \models \Diamond G) = 1\}$ .

### Algorithm

1. Make all states in  $G$  absorbing yielding  $\mathcal{D}[G]$ .
2. Determine  $S \setminus Pre^*(S \setminus Pre^*(G))$  by a graph analysis:
  - 2.1 do a backward search from  $G$  in  $\mathcal{D}[G]$  to determine  $Pre^*(G)$ .
  - 2.2 followed by a backward search from  $S \setminus Pre^*(G)$  in  $\mathcal{D}[G]$ .

This yields a time complexity which is linear in the size of the DTMC  $\mathcal{D}$ .

Thus a graph analysis suffices. No inspection of the probabilities is needed.

## Repeated reachability

### Almost sure repeated reachability theorem

For finite DTMC with state space  $S$ ,  $G \subseteq S$ , and  $s \in S$ :

$$Pr(s \models \Box \Diamond G) = 1 \quad \text{iff} \quad \text{for each BSCC } T \subseteq Post^*(s). T \cap G \neq \emptyset.$$

#### Proof:

Immediate consequence of the long-run theorem.

## Almost sure repeated reachability

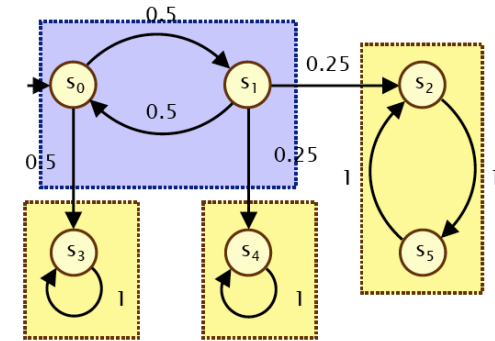
### Almost sure repeated reachability theorem

For finite DTMC with state space  $S$ ,  $G \subseteq S$ , and  $s \in S$ :

$$Pr(s \models \Box \Diamond G) = 1 \quad \text{iff} \quad \text{for each BSCC } T \subseteq Post^*(s). T \cap G \neq \emptyset.$$

Example:

$$B = \{s_3, s_4, s_5\}$$



## Almost sure persistence

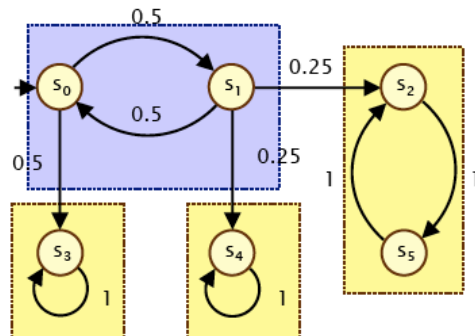
### Almost sure persistence theorem

For finite DTMC with state space  $S$ ,  $G \subseteq S$ , and  $s \in S$ :

$$Pr(s \models \Diamond \Box G) = 1 \quad \text{if and only if} \quad T \subseteq G \text{ for any BSCC } T \subseteq Post^*(s)$$

Example:

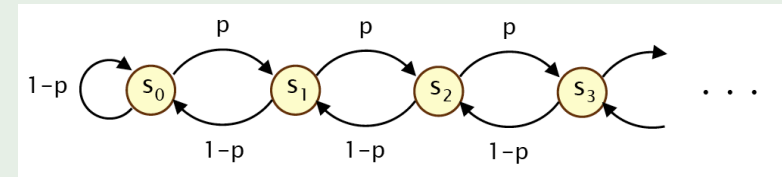
$$\{s_2, s_3, s_4, s_5\}$$



## A remark on infinite Markov chains

### Graph analysis for infinite DTMCs does not suffice!

Consider the following infinitely countable DTMC, known as **random walk**:



The value of rational probability  $p \in \mathbb{Q}$  **does** affect qualitative properties:

$$Pr(s \models \Diamond s_0) = \begin{cases} 1 & \text{if } p \leq \frac{1}{2} \\ < 1 & \text{if } p > \frac{1}{2} \end{cases} \quad \text{and}$$

$$Pr(s \models \Box \Diamond s_0) = \begin{cases} 1 & \text{if } p \leq \frac{1}{2} \\ 0 & \text{if } p > \frac{1}{2} \end{cases}$$

## Quantitative properties

### Quantitative repeated reachability theorem

For finite DTMC with state space  $S$ ,  $G \subseteq S$ , and  $s \in S$ :

$$Pr(s \models \Box \Diamond G) = Pr(s \models \Diamond U)$$

where  $U$  is the union of all BSCCs  $T$  with  $T \cap G \neq \emptyset$ .

### Quantitative persistence theorem

For finite DTMC with state space  $S$ ,  $G \subseteq S$ , and  $s \in S$ :

$$Pr(s \models \Diamond \Box G) = Pr(s \models \Diamond U)$$

where  $U$  is the union of all BSCCs  $T$  with  $T \subseteq G$ .

### Remark

Thus probabilities for  $\Box \Diamond G$  and  $\Diamond \Box G$  are reduced to **reachability probabilities**. These can be computed by solving a linear equation system.

## What remains

- ▶  $\Diamond \Box G$  and  $\Box \Diamond G$  are  $\omega$ -regular.
- ▶ Their likelihood can be reduced to reachability probabilities.
- ▶ How about arbitrary  $\omega$ -regular properties?
- ▶ Such as  $(\Diamond \Box F \wedge \Box \Diamond G)$  or  $\overline{F} U (\Diamond \Box G) \dots$
- ▶ Can they also be reduced to reachability probabilities? **Yes**, they can!

## Summary

- ▶ A finite DTMC almost surely ends up in a BSCC on the long run.
- ▶ Almost sure reachability = double backward search.
- ▶ Almost sure  $\Box \Diamond G$  and  $\Diamond \Box G$  properties can be checked by BSCC analysis and reachability.
- ▶ Probabilities for  $\Box \Diamond G$  and  $\Diamond \Box G$  reduce to reachability probabilities.

### Take-home message

For **finite** DTMCs, qualitative properties do only depend on their state graph and **not** on the transition probabilities! For infinite DTMCs, this does not hold.

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 **Verifying  $\omega$ -regular properties**
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation



## Paths and traces

### Paths

A **path** in DTMC  $\mathcal{D}$  is an infinite sequence of states  $s_0 s_1 s_2 \dots$  with  $\mathbf{P}(s_i, s_{i+1}) > 0$  for all  $i$ .

Let  $Paths(\mathcal{D})$  denote the set of paths in  $\mathcal{D}$ , and  $Paths^*(\mathcal{D})$  the set of finite prefixes thereof.

### Trace

The **trace** of path  $\pi = s_0 s_1 s_2 \dots$  is  $trace(\pi) = L(s_0) L(s_1) L(s_2) \dots$ . The trace of finite path  $\hat{\pi} = s_0 s_1 \dots s_n$  is  $trace(\hat{\pi}) = L(s_0) L(s_1) \dots L(s_n)$ .

The **set of traces** of a set  $\Pi$  of paths:  $trace(\Pi) = \{ trace(\pi) \mid \pi \in \Pi \}$ .

## Safety properties

### Safety property

LT property  $P_{safe}$  over  $AP$  is a **safety property** if for all  $\sigma \in (2^{AP})^\omega \setminus P_{safe}$  there exists a finite prefix  $\hat{\sigma}$  of  $\sigma$  such that:

$$P_{safe} \cap \underbrace{\left\{ \sigma' \in (2^{AP})^\omega \mid \hat{\sigma} \text{ is a prefix of } \sigma' \right\}}_{\text{all possible extensions of } \hat{\sigma}} = \emptyset.$$

Any such finite word  $\hat{\sigma}$  is called a **bad prefix** for  $P_{safe}$ .

### Regular safety property

A safety property is **regular** if its set of bad prefixes constitutes a regular language (over the alphabet  $2^{AP}$ ). Thus, the bad prefixes of a regular safety property can be represented by a finite-state automaton.

## LT properties

### Linear-time property

A **linear-time property** (LT property) over  $AP$  is a subset of  $(2^{AP})^\omega$ . An LT-property is thus a set of infinite traces over  $2^{AP}$ .

### Intuition

An LT-property gives the admissible behaviours of the DTMC at hand.

### Probability of LT properties

The **probability** for DTMC  $\mathcal{D}$  to exhibit a trace in  $P$  (over  $AP$ ) is:

$$Pr^{\mathcal{D}}(P) = Pr^{\mathcal{D}}\{ \pi \in Paths(\mathcal{D}) \mid trace(\pi) \in P \}.$$

For state  $s$  in  $\mathcal{D}$ , let  $Pr(s \models P) = Pr_s\{ \pi \in Paths(s) \mid trace(\pi) \in P \}$ .

We will later identify a rich set  $P$  of LT-properties—those that include all LTL formulas—for which  $\{ \pi \in Paths(\mathcal{D}) \mid trace(\pi) \in P \}$  is measurable.

## Probability of a regular safety property

Let  $\mathcal{A} = (Q, 2^{AP}, \delta, q_0, F)$  be a **deterministic finite-state automaton** (DFA) for the bad prefixes of regular safety property  $P_{safe}$ :

$$P_{safe} = \{ A_0 A_1 A_2 \dots \in (2^{AP})^\omega \mid \forall n \geq 0. A_0 A_1 \dots A_n \notin \mathcal{L}(\mathcal{A}) \}.$$

Assume  $\delta$  to be **total**, i.e.,  $\delta(q, A)$  is defined for each  $A \subseteq AP$  and each state  $q \in Q$ . Furthermore, let  $\mathcal{D} = (S, \mathbf{P}, \iota_{init}, AP, L)$  be a finite DTMC. Our interest is to compute the probability

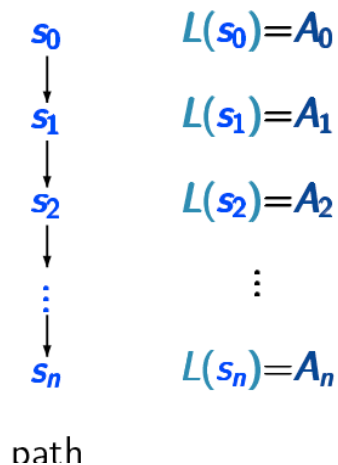
$$Pr^{\mathcal{D}}(P_{safe}) = 1 - \sum_{s \in S} \iota_{init}(s) \cdot Pr(s \models \mathcal{A}) \quad \text{where}$$

$$Pr(s \models \mathcal{A}) = Pr_s^{\mathcal{D}}\{ \pi \in Paths(s) \mid trace(\pi) \notin P_{safe} \}.$$

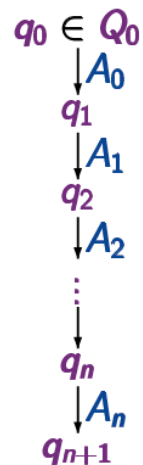
These probabilities can be obtained by considering a product of DTMC  $\mathcal{D}$  with DFA  $\mathcal{A}$ .

## Product construction: intuition

DTMC  $\mathcal{D}$   
with state space  $S$

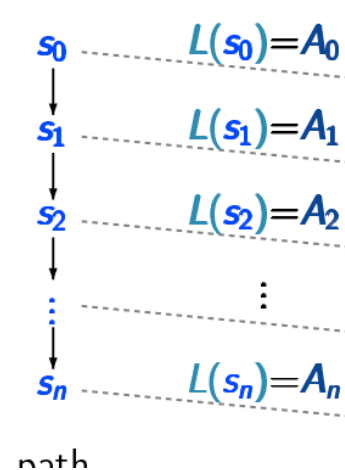


DRA  $\mathcal{A}$   
with state space  $Q$

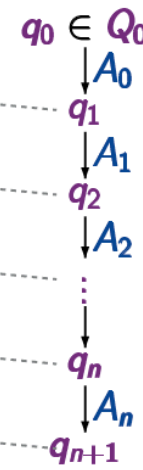


## Product construction: intuition

DTMC  $\mathcal{D}$   
with state space  $S$



DRA  $\mathcal{A}$   
with state space  $Q$



product  $\mathcal{D} \otimes \mathcal{A}$

## Product Markov chain

### Product Markov chain

Let  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  be a DTMC and  $\mathcal{A} = (Q, 2^{AP}, \delta, q_0, F)$  be a DFA. The **product**  $\mathcal{D} \otimes \mathcal{A}$  is the DTMC:

$$\mathcal{D} \otimes \mathcal{A} = (S \times Q, \mathbf{P}', \iota'_{\text{init}}, \{ \text{accept} \}, L')$$

where  $L'(\langle s, q \rangle) = \{ \text{accept} \}$  if  $q \in F$  and  $L'(\langle s, q \rangle) = \emptyset$  otherwise, and

$$\iota'_{\text{init}}(\langle s, q \rangle) = \begin{cases} \iota_{\text{init}}(s) & \text{if } q = \delta(q_0, L(s)) \\ 0 & \text{otherwise.} \end{cases}$$

The transition probabilities in  $\mathcal{D} \otimes \mathcal{A}$  are given by:

$$\mathbf{P}'(\langle s, q \rangle, \langle s', q' \rangle) = \begin{cases} \mathbf{P}(s, s') & \text{if } q' = \delta(q, L(s')) \\ 0 & \text{otherwise.} \end{cases}$$

## Quantitative analysis of regular safety properties

### Theorem for analysing regular safety properties

Let  $P_{\text{safe}}$  be a regular safety property,  $\mathcal{A}$  a DFA for the set of bad prefixes of  $P_{\text{safe}}$ ,  $\mathcal{D}$  a DTMC, and  $s$  a state in  $\mathcal{D}$ . Then:

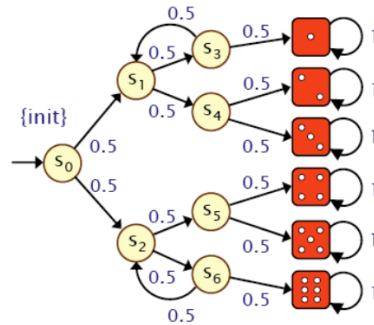
$$\begin{aligned} Pr^{\mathcal{D}}(s \models P_{\text{safe}}) &= Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \not\models \Diamond \text{accept}) \\ &= 1 - Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \Diamond \text{accept}) \end{aligned}$$

where  $q_s = \delta(q_0, L(s))$ .

### Remarks

1. For finite DTMCs,  $Pr^{\mathcal{D}}(s \models P_{\text{safe}})$  can thus be computed by determining **reachability probabilities** of **accept** states in  $\mathcal{D} \otimes \mathcal{A}$ . This amounts to solving a linear equation system.
2. For **qualitative** regular safety properties, i.e.,  $Pr^{\mathcal{D}}(s \models P_{\text{safe}}) > 0$  and  $Pr^{\mathcal{D}}(s \models P_{\text{safe}}) = 1$ , a graph analysis of  $\mathcal{D} \otimes \mathcal{A}$  suffices.

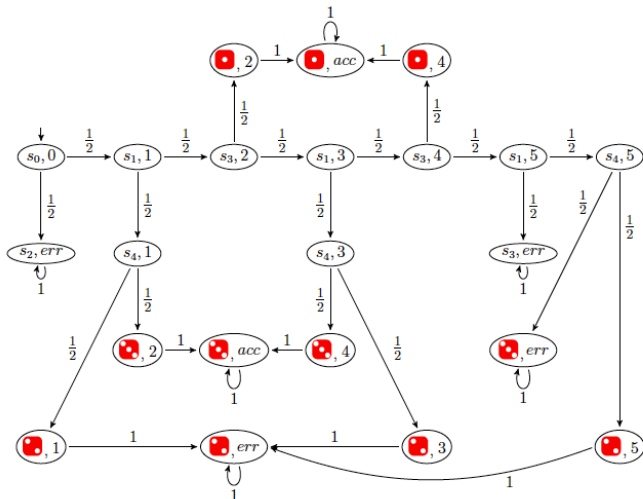
## Property of Knuth's die



### Property of Knuth's die

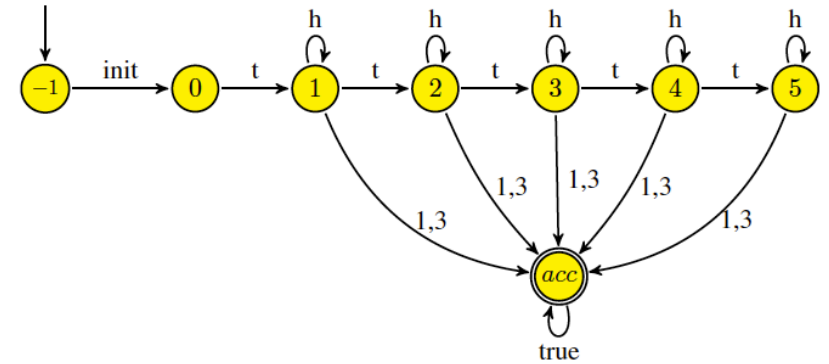
After initial tails, yield 1 or 3 but with maximally five time tails.

## Determining the property's probability



Reach probability of BSCC containing  $(\cdot, q_{acc})$  is  $\frac{1}{8} + \frac{1}{8} + \frac{1}{32} + \frac{1}{32} = \frac{5}{16}$ .

## Property as an automaton



After initial tails, yield 1 or 3 but with maximally five time tails

## $\omega$ -regular languages

### Infinite repetition of languages

Let  $\Sigma$  be a finite alphabet. For language  $\mathcal{L} \subseteq \Sigma^*$ , let  $\mathcal{L}^\omega$  be the set of words in  $\Sigma^* \cup \Sigma^\omega$  that arise from the infinite concatenation of (arbitrary) words in  $\Sigma$ , i.e.,

$$\mathcal{L}^\omega = \{w_1 w_2 w_3 \dots \mid w_i \in \mathcal{L}, i \geq 1\}.$$

The result is an  $\omega$ -language, i.e.,  $\mathcal{L} \subseteq \Sigma^*$ , provided that  $\mathcal{L} \subseteq \Sigma^+$ , i.e.,  $\varepsilon \notin \mathcal{L}$ .

### $\omega$ -regular expression

An  $\omega$ -regular expression  $G$  over the  $\Sigma$  has the form:

$$G = E_1.F_1^\omega + \dots + E_n.F_n^\omega$$

where  $n \geq 1$  and  $E_1, \dots, E_n, F_1, \dots, F_n$  are regular expressions over  $\Sigma$  such that  $\varepsilon \notin \mathcal{L}(F_i)$ , for all  $1 \leq i \leq n$ .

## Recall $\omega$ -regular expressions

### $\omega$ -regular expression

An  $\omega$ -regular expression  $G$  over the  $\Sigma$  has the form:

$$G = E_1.F_1^\omega + \dots + E_n.F_n^\omega$$

where  $n \geq 1$  and  $E_1, \dots, E_n, F_1, \dots, F_n$  are regular expressions over  $\Sigma$  such that  $\varepsilon \notin \mathcal{L}(F_i)$ , for all  $1 \leq i \leq n$ .

### Example

Let  $AP = \{a, b\}$ . Then some  $\omega$ -regular properties over  $AP$  are:

- ▶  $\Box a$ , i.e.,  $(\{a\} + \{a, b\})^\omega$ .
- ▶  $\Diamond a$ , i.e.,  $(\emptyset + \{b\})^* . (\{a\} + \{a, b\}) . (2^{AP})^\omega$ .
- ▶  $\Box \Diamond a$ , i.e.,  $((\emptyset + \{b\})^* . (\{a\} + \{a, b\}))^\omega$ .
- ▶  $\Diamond \Box a$ , i.e.,  $(2^{AP})^* . (\{a\} + \{a, b\})^\omega$ .

## LTL semantics

### LTL semantics

The LT-property induced by LTL formula  $\varphi$  over  $AP$  is:

$Words(\varphi) = \{\sigma \in (2^{AP})^\omega \mid \sigma \models \varphi\}$ , where  $\models$  is the smallest relation s.t.:

- $\sigma \models \text{true}$
- $\sigma \models a$  iff  $a \in A_0$  (i.e.,  $A_0 \models a$ )
- $\sigma \models \varphi_1 \wedge \varphi_2$  iff  $\sigma \models \varphi_1$  and  $\sigma \models \varphi_2$
- $\sigma \models \neg \varphi$  iff  $\sigma \not\models \varphi$
- $\sigma \models \bigcirc \varphi$  iff  $\sigma^1 = A_1 A_2 A_3 \dots \models \varphi$
- $\sigma \models \varphi_1 \mathbf{U} \varphi_2$  iff  $\exists j \geq 0. \sigma^j \models \varphi_2$  and  $\sigma^i \models \varphi_1, 0 \leq i < j$

for  $\sigma = A_0 A_1 A_2 \dots$  we have  $\sigma^i = A_i A_{i+1} A_{i+2} \dots$  is the suffix of  $\sigma$  from index  $i$  on.

## Linear temporal logic

### Linear Temporal Logic: Syntax

[Pnueli 1977]

LTL formulas over the set  $AP$  obey the grammar:

$$\varphi ::= a \mid \neg \varphi \mid \varphi_1 \wedge \varphi_2 \mid \bigcirc \varphi \mid \varphi_1 \mathbf{U} \varphi_2$$

where  $a \in AP$  and  $\varphi, \varphi_1$ , and  $\varphi_2$  are LTL formulas.

## Some facts about LTL

### LTL is $\omega$ -regular

For any LTL formula  $\varphi$ , the set  $Words(\varphi)$  is an  $\omega$ -regular language.

### LTL are DRA-definable

For any LTL formula  $\varphi$ , there exists a DRA  $\mathcal{A}$  such that  $\mathcal{L}_\omega = Words(\varphi)$  where the number of states in  $\mathcal{A}$  lies in  $2^{2^{|\varphi|}}$ .

## Deterministic Rabin automata

### Deterministic Rabin automaton

A *deterministic Rabin automaton* (DRA)  $\mathcal{A} = (Q, \Sigma, \delta, q_0, \mathcal{F})$  with

- ▶  $Q$  is a finite set of states
- ▶  $\Sigma$  is an alphabet
- ▶  $\delta : Q \times \Sigma \rightarrow Q$  is a transition function, and
- ▶  $q_0 \in Q$  is the initial state
- ▶  $\mathcal{F} = \{(L_i, K_i) \mid 0 < i \leq m\}$  with  $L_i, K_i \subseteq Q$ , is the *acceptance condition*

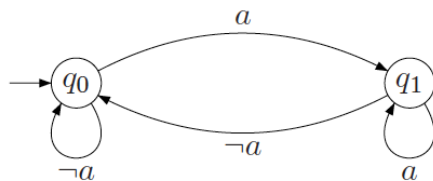
### Remark

The acceptance condition is a set of pairs of state sets. Recall that in Büchi automata this is simply a single set of states.

## Deterministic Rabin automaton: Example

### Acceptance condition

A run of a word in  $\Sigma^\omega$  on a DRA is *accepting* iff  $\bigvee_{0 < i \leq m} (\Diamond \Box \neg L_i \wedge \Box \Diamond K_i)$ .



For  $\mathcal{F} = \{(L, K)\}$  with  $L = \{q_0\}$  and  $K = \{q_1\}$ , this DRA accepts  $\Diamond \Box a$

Recall that there does not exist a *deterministic* Büchi automaton for  $\Diamond \Box a$ .

## When does a DRA accept an infinite word?

### Acceptance condition

A run of a word in  $\Sigma^\omega$  on a DRA is *accepting* if and only if:

for some  $(L_i, K_i) \in \mathcal{F}$ , the states in  $L_i$  are visited *finitely* often  
and (some of) the states in  $K_i$  are visited *infinitely* often

Stated in terms of an LTL formula:

$$\bigvee_{0 < i \leq m} (\Diamond \Box \neg L_i \wedge \Box \Diamond K_i)$$

A deterministic Büchi automaton is a DRA with acceptance condition  $\{(\emptyset, F)\}$ .

## Deterministic Rabin automata

### DRA are $\omega$ -regular

A language on infinite words is  $\omega$ -regular iff there exists a DRA that generates it.

- ▶ DRA are thus equally expressive as (generalized) Büchi automata.
- ▶ They are more expressive than deterministic Büchi automata.
- ▶ Any nondeterministic Büchi automata of  $n$  states can be converted to a DRA of size  $2^{\mathcal{O}(n \cdot \log n)}$ .

## Paths and traces

A *path* in DTMC  $\mathcal{D}$  is an infinite sequence of states  $s_0 s_1 s_2 \dots$  with  $P(s_i, s_{i+1}) > 0$  for all  $i$ .

### Trace

The *trace* of path  $\pi = s_0 s_1 s_2 \dots$  is  $\text{trace}(\pi) = L(s_0) L(s_1) L(s_2) \dots \in (2^{AP})^\omega$ .

### Probability of a DRA

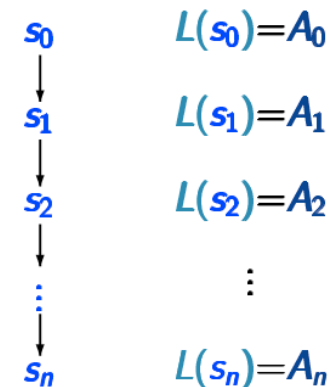
We consider DRAs over the alphabet  $\Sigma = 2^{AP}$ . Such DRAs *accept traces*. Our aim is to determine:

$$Pr(\mathcal{D} \models \mathcal{A}) = Pr\{\pi \in \text{Paths}(\mathcal{D}) \mid \text{trace}(\pi) \in \mathcal{L}_\omega(\mathcal{A})\}$$

(We will later see that this set is measurable.)

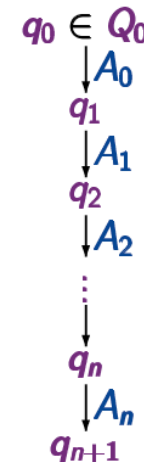
## Product construction: intuition

DTMC  $\mathcal{D}$   
with state space  $S$



nath

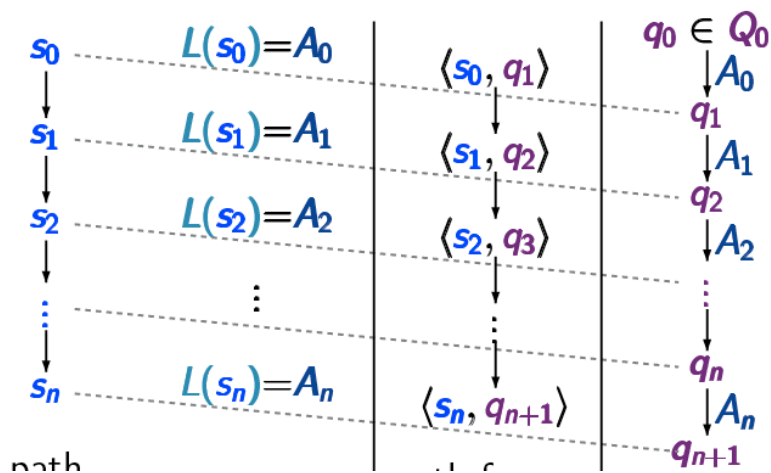
DRA  $\mathcal{A}$   
with state space  $Q$



## Product construction: intuition

DTMC  $\mathcal{D}$   
with state space  $S$

DRA  $\mathcal{A}$   
with state space  $Q$



product  $\mathcal{D} \otimes \mathcal{A}$

## Product Markov chain

### Product Markov chain

Let  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  be a DTMC and  $\mathcal{A} = (Q, 2^{AP}, \delta, q_0, \mathcal{F})$  be a DRA. The *product*  $\mathcal{D} \otimes \mathcal{A}$  is the DTMC:

$$\mathcal{D} \otimes \mathcal{A} = (S \times Q, \mathbf{P}', \iota'_{\text{init}}, 2^Q, L')$$

where  $L_i, K_j \in L'(\langle s, q \rangle)$  iff  $q \in L_i$  or  $q \in K_j$  and

$$\iota'_{\text{init}}(\langle s, q \rangle) = \begin{cases} \iota_{\text{init}}(s) & \text{if } q = \delta(q_0, L(s)) \\ 0 & \text{otherwise.} \end{cases}$$

The transition probabilities in  $\mathcal{D} \otimes \mathcal{A}$  are given by:

$$\mathbf{P}'(\langle s, q \rangle, \langle s', q' \rangle) = \begin{cases} \mathbf{P}(s, s') & \text{if } q' = \delta(q, L(s')) \\ 0 & \text{otherwise.} \end{cases}$$

## Verifying DRA properties

### Accepting BSCC

A BSCC  $T$  in  $\mathcal{D} \otimes \mathcal{A}$  is **accepting** iff there exists some index  $i \in \{1, \dots, m\}$  such that:

$$T \cap (S \times L_i) = \emptyset \quad \text{and} \quad T \cap (S \times K_i) \neq \emptyset.$$

Thus, once such an accepting BSCC  $T$  is reached in  $\mathcal{D} \otimes \mathcal{A}$ , the acceptance criterion for the DRA  $\mathcal{A}$  is fulfilled almost surely.

### DRA probabilities = reachability probabilities

Let  $\mathcal{D}$  be a finite DTMC,  $s$  a state in  $\mathcal{D}$ ,  $\mathcal{A}$  a DRA, and let  $U$  be the union of all **accepting** BSCCs in  $\mathcal{D} \otimes \mathcal{A}$ . Then:

$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \Diamond U) \quad \text{where} \quad q_s = \delta(q_0, L(s)).$$

## Measurability

### Measurability theorem for $\omega$ -regular properties

[Vardi 1985]

For any DTMC  $\mathcal{D}$  and DRA  $\mathcal{A}$  the set

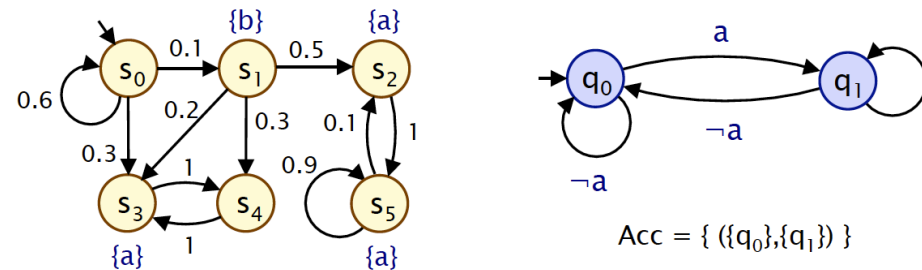
$$\{\pi \in \text{Paths}(\mathcal{D}) \mid \text{trace}(\pi) \in \mathcal{L}_\omega(\mathcal{A})\}$$

is measurable.

### Proof (sketch)

Let DRA  $\mathcal{A}$  with accept sets  $\{(L_1, K_1), \dots, (L_m, K_m)\}$ . Let  $\varphi_i = \Diamond \Box \neg L_i \wedge \Box \Diamond K_i$  and  $\Pi_i$  the set of paths satisfying  $\varphi_i$ . Then  $\Pi = \Pi_1 \cup \dots \cup \Pi_k$ . In addition,  $\Pi_i = \Pi_i^{\Diamond \Box} \cap \Pi_i^{\Box \Diamond}$  where  $\Pi_i^{\Diamond \Box}$  is the set of paths  $\pi$  in  $\mathcal{D}$  such that  $\pi^+ \models \Diamond \Box \neg L_i$ , and  $\Pi_i^{\Box \Diamond}$  is the set of paths  $\pi$  in  $\mathcal{D}$  such that  $\pi^+ \models \Box \Diamond K_i$ . It remains to show that  $\Pi_i^{\Diamond \Box}$  and  $\Pi_i^{\Box \Diamond}$  are measurable. This goes along the same lines as proving that  $\Diamond \Box G$  and  $\Box \Diamond G$  are measurable.

## Example: verifying a DTMC versus a DRA



Single accepting BSCC:  $\{\langle s_2, q_1 \rangle, \langle s_5, q_1 \rangle\}$ .

Reachability probability is  $\frac{1}{2} \cdot \frac{1}{10} \cdot \sum_{k=0}^{\infty} \left(\frac{3}{5}\right)^k = \frac{1}{8}$ .

## Probabilities for LTL formulas

### LTL are DRA-definable

For any LTL formula  $\varphi$ , there exists a DRA  $\mathcal{A}$  such that  $\mathcal{L}_\omega(\mathcal{A}) = \text{Words}(\varphi)$  where the number of states in  $\mathcal{A}$  lies in  $2^{2|\varphi|}$ .

### Complexity of LTL model checking

[Vardi 1985]

The **qualitative** model-checking problem for finite DTMCs against LTL formula  $\varphi$  is PSPACE-complete, i.e., verifying whether  $Pr(s \models \varphi) > 0$  or  $Pr(s \models \varphi) = 1$  is PSPACE-complete.

Qualitative LTL model checking of Markov chains falls in the same complexity class as LTL model checking of Kripke structures.

## Summary

### Summary

- ▶ Verifying a DTMC  $\mathcal{D}$  against a DFA  $\mathcal{A}$ , i.e., determining  $Pr(\mathcal{D} \models \mathcal{A})$ , amounts to computing reachability probabilities of accept states in  $\mathcal{D} \otimes \mathcal{A}$ .
- ▶ For DBA objectives, the probability of infinitely often visiting an accept state in  $\mathcal{D} \otimes \mathcal{A}$ .
- ▶ DBA are strictly less powerful than  $\omega$ -regular languages.
- ▶ Deterministic Rabin automata are as expressive as  $\omega$ -regular languages.
- ▶ Verifying DTMC  $\mathcal{D}$  against DRA  $\mathcal{A}$  amounts to computing reachability probabilities of accepting BSCCs in  $\mathcal{D} \otimes \mathcal{A}$ .

### Take-home message

Model checking a DTMC against various automata models reduces to computing reachability probabilities in a product.

## Probabilistic Computation Tree Logic

- ▶ PCTL is a language for formally specifying properties over DTMCs.
- ▶ It is a branching-time temporal logic based on CTL.
- ▶ Formula interpretation is Boolean: a state satisfies a formula or not.
- ▶ The main operator is  $\mathbb{P}_J(\varphi)$ 
  - ▶ where  $\varphi$  constrains the set of paths, and
  - ▶  $J$  is a threshold on the probability.
  - ▶ it is the probabilistic counterpart of  $\exists$  and  $\forall$  path-quantifiers in CTL.

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

## PCTL syntax

[Hansson & Jonsson, 1994]

### Probabilistic Computation Tree Logic: Syntax

PCTL consists of state- and path-formulas.

- ▶ PCTL *state formulas* over the set  $AP$  obey the grammar:

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg\Phi \mid \mathbb{P}_J(\varphi)$$

where  $a \in AP$ ,  $\varphi$  is a path formula and  $J \subseteq [0, 1]$ ,  $J \neq \emptyset$  is a non-empty interval.

- ▶ PCTL *path formulae* are formed according to the following grammar:

$$\varphi ::= \bigcirc\Phi \mid \Phi_1 \cup \Phi_2 \mid \Phi_1 \cup^{\leq n} \Phi_2$$

where  $\Phi$ ,  $\Phi_1$ , and  $\Phi_2$  are state formulae and  $n \in \mathbb{N}$ .

Abbreviate  $\mathbb{P}_{[0,0.5]}(\varphi)$  by  $\mathbb{P}_{\leq 0.5}(\varphi)$  and  $\mathbb{P}_{[0,1]}(\varphi)$  by  $\mathbb{P}_{>0}(\varphi)$ .



## Probabilistic Computation Tree Logic

- PCTL *state formulas* over the set  $AP$  obey the grammar:

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg\Phi \mid \mathbb{P}_J(\varphi)$$

where  $a \in AP$ ,  $\varphi$  is a path formula and  $J \subseteq [0, 1]$ ,  $J \neq \emptyset$  is a non-empty interval.

- PCTL *path formulae* are formed according to the following grammar:

$$\varphi ::= \bigcirc\Phi \mid \Phi_1 \cup \Phi_2 \mid \Phi_1 \cup^{\leq n} \Phi_2 \quad \text{where } n \in \mathbb{N}.$$

### Intuitive semantics

- $s_0 s_1 s_2 \dots \models \Phi \cup^{\leq n} \Psi$  if  $\Phi$  holds until  $\Psi$  holds within  $n$  steps.
- $s \models \mathbb{P}_J(\varphi)$  if probability that paths starting in  $s$  fulfill  $\varphi$  lies in  $J$ .

## Derived operators

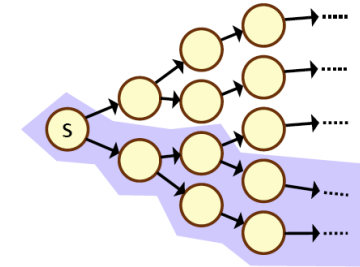
$$\Diamond\Phi = \text{true} \cup \Phi$$

$$\Diamond^{\leq n}\Phi = \text{true} \cup^{\leq n} \Phi$$

$$\mathbb{P}_{\leq p}(\Box\Phi) = \mathbb{P}_{>1-p}(\Diamond\neg\Phi)$$

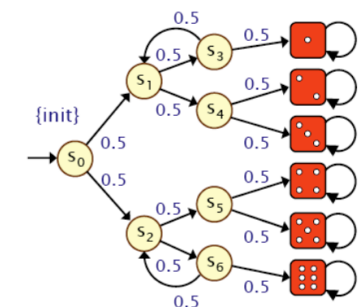
$$\mathbb{P}_{(p,q)}(\Box^{\leq n}\Phi) = \mathbb{P}_{[1-q,1-p]}(\Diamond^{\leq n}\neg\Phi)$$

## Semantics of $\mathbb{P}$ -operator



- $s \models \mathbb{P}_J(\varphi)$  if:
  - the probability of all paths starting in  $s$  fulfilling  $\varphi$  lies in  $J$ .
- Example:  $s \models \mathbb{P}_{>\frac{1}{2}}(\Diamond a)$  if
  - the probability to reach an  $a$ -labeled state from  $s$  exceeds  $\frac{1}{2}$ .
- Formally:
  - $s \models \mathbb{P}_J(\varphi)$  if and only if  $Pr_s\{\pi \in Paths(s) \mid \pi \models \varphi\} \in J$ .

## Correctness of Knuth's die



### Correctness of Knuth's die

$$\mathbb{P}_{=\frac{1}{6}}(\Diamond 1) \wedge \mathbb{P}_{=\frac{1}{6}}(\Diamond 2) \wedge \mathbb{P}_{=\frac{1}{6}}(\Diamond 3) \wedge \mathbb{P}_{=\frac{1}{6}}(\Diamond 4) \wedge \mathbb{P}_{=\frac{1}{6}}(\Diamond 5) \wedge \mathbb{P}_{=\frac{1}{6}}(\Diamond 6)$$

## Measurability

### PCTL measurability

For any PCTL path formula  $\varphi$  and state  $s$ ,  $\{\pi \in \text{Paths}(s) \mid \pi \models \varphi\}$  is measurable.

### Proof (sketch):

Three cases:

1.  $\bigcirc \Phi$ :
  - ▶ cylinder sets constructed from paths of length one.
2.  $\Phi \cup^{\leq n} \Psi$ :
  - ▶ (finite number of) cylinder sets from paths of length at most  $n$ .
3.  $\Phi \cup \Psi$ :
  - ▶ countable union of paths satisfying  $\Phi \cup^{\leq n} \Psi$  for all  $n \geq 0$ .

## Core model checking algorithm

### Probabilistic operator $\mathbb{P}$

In order to determine whether  $s \in \text{Sat}(\mathbb{P}_J(\varphi))$ , the probability  $Pr(s \models \varphi)$  for the event specified by  $\varphi$  needs to be established. Then

$$\text{Sat}(\mathbb{P}_J(\varphi)) = \{s \in S \mid Pr(s \models \varphi) \in J\}.$$

Let us consider the computation of  $Pr(s \models \varphi)$  for all possible  $\varphi$ .

## PCTL model checking

### PCTL model checking problem

**Input:** a finite DTMC  $\mathcal{D} = (S, \mathbf{P}, \ell_{\text{init}}, AP, L)$ , state  $s \in S$ , and PCTL state formula  $\Phi$

**Output:** yes, if  $s \models \Phi$ ; no, otherwise.

### Basic algorithm

In order to check whether  $s \models \Phi$  do:

1. Compute the **satisfaction set**  $\text{Sat}(\Phi) = \{s \in S \mid s \models \Phi\}$ .
2. This is done **recursively** by a bottom-up traversal of  $\Phi$ 's parse tree.
  - ▶ The nodes of the parse tree represent the subformulae of  $\Phi$ .
  - ▶ For each node, i.e., for each subformula  $\Psi$  of  $\Phi$ , determine  $\text{Sat}(\Psi)$ .
  - ▶ Determine  $\text{Sat}(\Psi)$  as function of the satisfaction sets of its children:  
e.g.,  $\text{Sat}(\Psi_1 \wedge \Psi_2) = \text{Sat}(\Psi_1) \cap \text{Sat}(\Psi_2)$  and  $\text{Sat}(\neg \Psi) = S \setminus \text{Sat}(\Psi)$ .
3. Check whether state  $s$  belongs to  $\text{Sat}(\Phi)$ .

## The next-step operator

Recall that:  $s \models \mathbb{P}_J(\bigcirc \Phi)$  if and only if  $Pr(s \models \bigcirc \Phi) \in J$ .

### Lemma

$$Pr(s \models \bigcirc \Phi) = \sum_{s' \in \text{Sat}(\Phi)} \mathbf{P}(s, s').$$

### Algorithm

Considering the above equation for all states simultaneously yields:

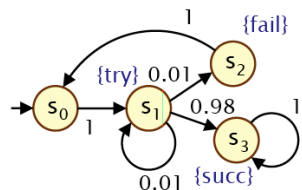
$$(Pr(s \models \bigcirc \Phi))_{s \in S} = \mathbf{P} \cdot \mathbf{b}_\Phi$$

with  $\mathbf{b}_\Phi$  the characteristic vector of  $\text{Sat}(\Phi)$ , i.e.,  $b_\Phi(s) = 1$  iff  $s \in \text{Sat}(\Phi)$ .

Checking the next-step operator reduces to a single matrix-vector multiplication.

## Example

Consider DTMC:



and PCTL-formula:

$$\mathbb{P}_{\geq 0.9}(\bigcirc(\neg try \vee succ))$$

1.  $Sat(\neg try \vee succ) = (S \setminus Sat(try)) \cup Sat(succ) = \{s_0, s_2, s_3\}$
2. We know:  $(Pr(s \models \bigcirc \Phi))_{s \in S} = \mathbf{P} \cdot \mathbf{b}_\Phi$  where  $\Phi = \neg try \vee succ$
3. Applying that to this example yields:

$$(Pr(s \models \bigcirc \Phi))_{s \in S} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0.01 & 0.01 & 0.98 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0.99 \\ 1 \\ 1 \end{pmatrix}$$

4. Thus:  $Sat(\mathbb{P}_{\geq 0.9}(\bigcirc(\neg try \vee succ))) = \{s_1, s_2, s_3\}$ .

## Time complexity

### Time complexity of PCTL model checking

For finite DTMC  $\mathcal{D}$  and PCTL state-formula  $\Phi$ , the PCTL model-checking problem can be solved in time

$$\mathcal{O}(\text{poly}(\text{size}(\mathcal{D})) \cdot n_{\max} \cdot |\Phi|).$$

### Proof (sketch)

1. For each node in the parse tree, a model-checking is performed; this yields a linear complexity in  $|\Phi|$ .
2. The worst-case operator is (unbounded) until.
  - 2.1 Determining  $S_{=0}$  and  $S_{=1}$  can be done in linear time.
  - 2.2 Direct methods to solve linear equation systems are in  $\Theta(|S|^3)$ .
3. Strictly speaking,  $U^{\leq n}$  could be more expensive for large  $n$ .  
But it remains polynomial, and  $n$  is small in practice.

## Time complexity

Let  $|\Phi|$  be the size of  $\Phi$ , i.e., the number of logical and temporal operators in  $\Phi$ .

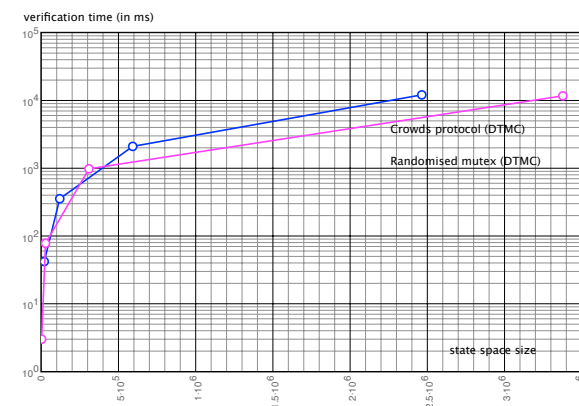
### Time complexity of PCTL model checking

For finite DTMC  $\mathcal{D}$  and PCTL state-formula  $\Phi$ , the PCTL model-checking problem can be solved in time

$$\mathcal{O}(\text{poly}(\text{size}(\mathcal{D})) \cdot n_{\max} \cdot |\Phi|)$$

where  $n_{\max} = \max\{n \mid \Psi_1 U^{\leq n} \Psi_2 \text{ occurs in } \Phi\}$  with  $n_{\max} = 1$  if  $\Phi$  does not contain a bounded until-operator.

## Some practical verification times



- command-line tool MRMC ran on a Pentium 4, 2.66 GHz, 1 GB RAM laptop.
- PCTL formula  $\mathbb{P}_{\leq p}(\diamond obs)$  where  $obs$  holds when the sender's id is detected.

## Summary

- ▶ PCTL is a variant of CTL with operator  $\mathbb{P}_J(\varphi)$ .
- ▶ Sets of paths fulfilling PCTL path-formula  $\varphi$  are measurable.
- ▶ PCTL model checking is performed by a recursive descent over  $\Phi$ .
- ▶ The next operator amounts to a single matrix-vector multiplication.
- ▶ The bounded-until operator  $U^{\leq n}$  amounts to  $n$  matrix-vector multiplications.
- ▶ The until-operator amounts to solving a linear equation system.
- ▶ The worst-case time complexity is polynomial in the size of the DTMC and linear in the size of the formula.

## Qualitative PCTL

### Qualitative PCTL

State formulae in the *qualitative fragment* of PCTL (over  $AP$ ):

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg\Phi \mid \mathbb{P}_{>0}(\varphi) \mid \mathbb{P}_{=1}(\varphi)$$

where  $a \in AP$ , and  $\varphi$  is a path formula formed according to the grammar:

$$\varphi ::= \bigcirc \Phi \mid \Phi_1 U \Phi_2.$$

### Remark

The probability bounds  $= 0$  and  $< 1$  can be derived:

$$\mathbb{P}_{=0}(\varphi) \equiv \neg\mathbb{P}_{>0}(\varphi) \quad \text{and} \quad \mathbb{P}_{<1}(\varphi) \equiv \neg\mathbb{P}_{=1}(\varphi)$$

So, in qualitative PCTL, there is no bounded until, and only  $> 0$ ,  $= 0$ ,  $> 1$  and  $= 1$  thresholds.

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

## Qualitative PCTL

### Qualitative PCTL

State formulae in the *qualitative fragment* of PCTL (over  $AP$ ):

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg\Phi \mid \mathbb{P}_{>0}(\varphi) \mid \mathbb{P}_{=1}(\varphi)$$

where  $a \in AP$ , and  $\varphi$  is a path formula formed according to the grammar:

$$\varphi ::= \bigcirc \Phi \mid \Phi_1 U \Phi_2.$$

### Examples

$\mathbb{P}_{=1}(\Diamond\mathbb{P}_{>0}(\bigcirc a))$  and  $\mathbb{P}_{<1}(\mathbb{P}_{>0}(\Diamond a) U b)$  are qualitative PCTL formulas.

## CTL versus qualitative PCTL

### Equivalence of PCTL and CTL Formulae

The PCTL formula  $\Phi$  is *equivalent* to the CTL formula  $\Psi$ , denoted  $\Phi \equiv \Psi$ , if  $Sat(\Phi) = Sat(\Psi)$  for each DTMC  $\mathcal{D}$ .

### Example

The simplest such cases are path formulae involving the next-step operator:

$$\mathbb{P}_{=1}(\bigcirc a) \equiv \forall \bigcirc a$$

$$\mathbb{P}_{>0}(\bigcirc a) \equiv \exists \bigcirc a$$

And for  $\exists \Diamond$  and  $\forall \Box$  we have:

$$\mathbb{P}_{>0}(\Diamond a) \equiv \exists \Diamond a$$

$$\mathbb{P}_{=1}(\Box a) \equiv \forall \Box a.$$

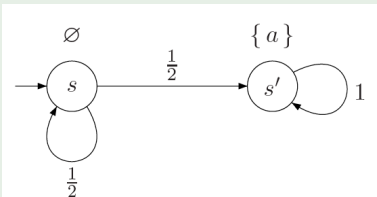
## CTL versus qualitative PCTL

$$(1) \mathbb{P}_{>0}(\Diamond a) \equiv \exists \Diamond a \quad \text{and} \quad (2) \mathbb{P}_{=1}(\Box a) \equiv \forall \Box a.$$

$$(3) \mathbb{P}_{>0}(\Box a) \not\equiv \exists \Box a \quad \text{and} \quad (4) \mathbb{P}_{=1}(\Diamond a) \not\equiv \forall \Diamond a.$$

### Example

Consider the second statement (4). Let  $s$  be a state in a (possibly infinite) DTMC. Then:  $s \models \forall \Diamond a$  implies  $s \models \mathbb{P}_{=1}(\Diamond a)$ . The reverse direction, however, does not hold. Consider the example DTMC:



$s \models \mathbb{P}_{=1}(\Diamond a)$  as the probability of path  $s^\omega$  is zero. However, the path  $s^\omega$  is possible and violates  $\Diamond a$ . Thus,  $s \not\models \forall \Diamond a$ .

Statement (3) follows by duality.

## CTL versus qualitative PCTL

$$(1) \mathbb{P}_{>0}(\Diamond a) \equiv \exists \Diamond a \quad \text{and} \quad (2) \mathbb{P}_{=1}(\Box a) \equiv \forall \Box a.$$

### Proof:

(1) Consider the first statement.

$\Rightarrow$  Assume  $s \models \mathbb{P}_{>0}(\Diamond a)$ . By the PCTL semantics,  $Pr(s \models \Diamond a) > 0$ . Thus,  $\{\pi \in Paths(s) \mid \pi \models \Diamond a\} \neq \emptyset$ , and hence,  $s \models \exists \Diamond a$ .

$\Leftarrow$  Assume  $s \models \exists \Diamond a$ , i.e., there is a finite path  $\hat{\pi} = s_0 s_1 \dots s_n$  with  $s_0 = s$  and  $s_n \models a$ . It follows that all paths in the cylinder set  $Cyl(\hat{\pi})$  fulfill  $\Diamond a$ . Thus:

$$Pr(s \models \Diamond a) \geq Pr_s(Cyl(s_0 s_1 \dots s_n)) = P(s_0 s_1 \dots s_n) > 0.$$

So,  $s \models \mathbb{P}_{>0}(\Diamond a)$ .

(2) The second statement follows by duality.

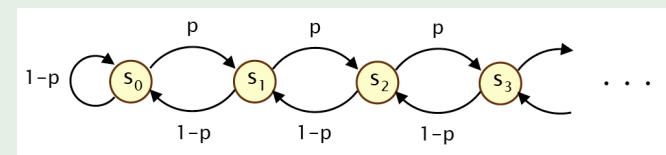
## Almost-sure-reachability not in CTL

### Almost-sure-reachability not in CTL

1. There is no CTL formula that is equivalent to  $\mathbb{P}_{=1}(\Diamond a)$ .
2. There is no CTL formula that is equivalent to  $\mathbb{P}_{>0}(\Box a)$ .

### Proof:

We provide the proof of 1.; 2. follows by duality:  $\mathbb{P}_{=1}(\Diamond a) \equiv \neg \mathbb{P}_{>0}(\Box \neg a)$ . By contraposition. Assume  $\Phi \equiv \mathbb{P}_{=1}(\Diamond a)$ . Consider the infinite DTMC  $\mathcal{D}_p$ :



The value of  $p$  **does** affect reachability:  $Pr(s \models \Diamond s_0) = \begin{cases} 1 & \text{if } p \leq \frac{1}{2} \\ < 1 & \text{if } p > \frac{1}{2} \end{cases}$

## Almost-sure-reachability not in CTL

There is no CTL formula that is equivalent to  $\mathbb{P}_{=1}(\Diamond a)$ .

### Proof:

$$\text{We have: } \Pr(s \models \Diamond s_0) = \begin{cases} 1 & \text{if } p \leq \frac{1}{2} \\ < 1 & \text{if } p > \frac{1}{2} \end{cases}$$

Thus, in  $\mathcal{D}_{\frac{1}{4}}$  we have  $s \models \mathbb{P}_{=1}(\Diamond s_0)$  for all states  $s$ , while in  $\mathcal{D}_{\frac{3}{4}}$ , e.g.,  $s_1 \not\models \mathbb{P}_{=1}(\Diamond s_0)$ . Hence:  $s_1 \in \text{Sat}_{\mathcal{D}_{\frac{1}{4}}}(\mathbb{P}_{=1}(\Diamond s_0))$  but  $s_1 \notin \text{Sat}_{\mathcal{D}_{\frac{3}{4}}}(\mathbb{P}_{=1}(\Diamond s_0))$ . For CTL-formula  $\Phi$  —by assumption  $\Phi \equiv \mathbb{P}_{=1}(\Diamond s_0)$ — we have:

$$\text{Sat}_{\mathcal{D}_{\frac{1}{4}}}(\Phi) = \text{Sat}_{\mathcal{D}_{\frac{3}{4}}}(\Phi).$$

Hence, state  $s_1$  either fulfills the CTL formula  $\Phi$  in both DTMCs or in none of them. This, however, contradicts  $\Phi \equiv \mathbb{P}_{=1}(\Diamond s_0)$ .

## Fair CTL

### Fair paths

In **fair** CTL, path formulas are interpreted over **fair** infinite paths, i.e., paths  $\pi$  that satisfy

$$\text{fair} = \bigwedge_{s \in S} \bigwedge_{t \in \text{Post}(s)} (\Box \Diamond s \rightarrow \Box \Diamond t).$$

A path  $\pi$  such that  $\pi \models \text{fair}$  is called **fair**. Let  $\text{Paths}_{\text{fair}}(s)$  be the set of fair paths starting in  $s$ .

### Fair CTL semantics

The **fair semantics** of CTL is defined by the satisfaction  $\models_{\text{fair}}$  which is defined as  $\models$  for the CTL semantics, except that:

$$\begin{aligned} s \models_{\text{fair}} \exists \varphi & \text{ iff there exists } \pi \in \text{Paths}_{\text{fair}}(s). \pi \models_{\text{fair}} \varphi \\ s \models_{\text{fair}} \forall \varphi & \text{ iff for all } \pi \in \text{Paths}_{\text{fair}}(s). \pi \models_{\text{fair}} \varphi. \end{aligned}$$

## $\forall \Diamond$ is not expressible in qualitative PCTL

1. There is no qualitative PCTL formula that is equivalent to  $\forall \Diamond a$ .
2. There is no qualitative PCTL formula that is equivalent to  $\exists \Box a$ .

## Fairness theorem

### Qualitative PCTL versus fair CTL theorem

Let  $s$  be an arbitrary state in a finite DTMC. Then:

$$\begin{aligned} s \models \mathbb{P}_{=1}(\Diamond a) & \text{ iff } s \models_{\text{fair}} \forall \Diamond a \\ s \models \mathbb{P}_{>0}(\Box a) & \text{ iff } s \models_{\text{fair}} \exists \Box a \\ s \models \mathbb{P}_{=1}(a \cup b) & \text{ iff } s \models_{\text{fair}} \forall (a \cup b) \\ s \models \mathbb{P}_{>0}(a \cup b) & \text{ iff } s \models_{\text{fair}} \exists (a \cup b) \end{aligned}$$

### Comparable expressiveness

Qualitative PCTL and fair CTL are equally expressive.

## Almost sure repeated reachability

### Almost sure repeated reachability is PCTL-definable

For finite DTMC  $\mathcal{D}$ , state  $s \in S$  and  $G \subseteq S$ :

$$s \models \mathbb{P}_{=1}(\Box \mathbb{P}_{=1}(\Diamond G)) \quad \text{iff} \quad \Pr_s\{\pi \in \text{Paths}(s) \mid \pi \models \Box \Diamond G\} = 1.$$

We abbreviate  $\mathbb{P}_{=1}(\Box \mathbb{P}_{=1}(\Diamond G))$  by  $\mathbb{P}_{=1}(\Box \Diamond G)$ .

#### Remark:

For CTL, universal repeated reachability properties can be formalized by the combination of the modalities  $\forall\Box$  and  $\forall\Diamond$ :

$$s \models \forall\Diamond\forall\Diamond G \quad \text{iff} \quad \pi \models \Box \Diamond G \text{ for all } \pi \in \text{Paths}(s).$$

## Almost sure persistence

### Almost sure persistence is PCTL-definable

For finite DTMC  $\mathcal{D}$ , state  $s \in S$  and  $G \subseteq S$ :

$$s \models \mathbb{P}_{=1}(\Diamond \mathbb{P}_{=1}(\Box G)) \quad \text{iff} \quad \Pr_s\{\pi \in \text{Paths}(s) \mid \pi \models \Diamond \Box G\} = 1.$$

We abbreviate  $\mathbb{P}_{=1}(\Diamond \mathbb{P}_{=1}(\Box G))$  by  $\mathbb{P}_{=1}(\Diamond \Box G)$ .

#### Remark:

Note that  $\forall\Diamond\Box G$  is not CTL-definable.  $\Diamond\Box G$  is a well-known example formula in LTL that cannot be expressed in CTL. But by the above theorem it can be expressed in PCTL.

## Repeated reachability probabilities

### Repeated reachability probabilities are PCTL-definable

For finite DTMC  $\mathcal{D}$ , state  $s \in S$ ,  $G \subseteq S$  and interval  $J \subseteq [0, 1]$  we have:

$$s \models \mathbb{P}_J(\underbrace{\Diamond \mathbb{P}_{=1}(\Box \mathbb{P}_{=1}(\Diamond G))}_{=\mathbb{P}_J(\Box \Diamond G)}) \quad \text{if and only if} \quad \Pr(s \models \Box \Diamond G) \in J.$$

#### Remark:

By the above theorem,  $\mathbb{P}_{>0}(\Box \Diamond G)$  is PCTL definable. Note that  $\exists\Box \Diamond G$  is not CTL-definable (but definable in a combination of CTL and LTL, called CTL\*).

## Persistence probabilities

### Persistence probabilities are PCTL-definable

For finite DTMC  $\mathcal{D}$ , state  $s \in S$ ,  $G \subseteq S$  and interval  $J \subseteq [0, 1]$  we have:

$$s \models \mathbb{P}_J(\underbrace{\Diamond \mathbb{P}_{=1}(\Box G)}_{=\mathbb{P}_J(\Diamond \Box G)}) \quad \text{if and only if} \quad \Pr(s \models \Diamond \Box G) \in J.$$

#### Proof:

Left as an exercise. Hint: use the long run theorem.

## Summary

- ▶ Qualitative PCTL only allow the probability bounds  $> 0$  and  $= 1$ .
- ▶ There is no CTL formula that is equivalent to  $\mathbb{P}_{=1}(\Diamond a)$ .
- ▶ There is no PCTL formula that is equivalent to  $\forall \Box a$ .
- ▶ These results do not apply to finite DTMCs.
- ▶  $\mathbb{P}_{=1}(\Diamond a)$  and  $\forall \Diamond a$  are equivalent under fairness.
- ▶ Repeated reachability probabilities are PCTL definable.

### Take-home messages

Qualitative PCTL and CTL have incomparable expressiveness. Qualitative and fair CTL are equally expressive. Repeated reachability and persistence probabilities are PCTL definable. Their qualitative counterparts are not expressible in CTL.

## Probabilistic bisimulation: intuition

### Intuition

- ▶ Strong bisimulation is used to **compare** labeled transition systems.
- ▶ Strongly bisimilar states exhibit the same step-wise behaviour.
- ▶ Our aim: adapt bisimulation to discrete-time Markov chains.
- ▶ This yields a probabilistic variant of strong bisimulation.

- ▶ When do two DTMC states exhibit the same step-wise behaviour?
- ▶ **Key: if their transition probability for each equivalence class coincides.**

## Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities
- 4 Qualitative reachability and all that
- 5 Verifying  $\omega$ -regular properties
- 6 Verifying probabilistic CTL
- 7 Expressiveness of probabilistic CTL
- 8 Probabilistic bisimulation

## Probabilistic bisimulation

### Probabilistic bisimulation

[Larsen &amp; Skou, 1989]

Let  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  be a DTMC and  $R \subseteq S \times S$  an **equivalence**. Then:  $R$  is a **probabilistic bisimulation** on  $S$  if for any  $(s, t) \in R$ :

1.  $L(s) = L(t)$ , and
2.  $\mathbf{P}(s, C) = \mathbf{P}(t, C)$  for all equivalence classes  $C \in S/R$

where  $\mathbf{P}(s, C) = \sum_{s' \in C} \mathbf{P}(s, s')$ .

For states in  $R$ , the probability of moving by a single transition to some equivalence class is equal.

### Probabilistic bisimilarity

Let  $\mathcal{D}$  be a DTMC and  $s, t$  states in  $\mathcal{D}$ . Then:  $s$  is **probabilistically bisimilar** to  $t$ , denoted  $s \sim_p t$ , if there **exists** a probabilistic bisimulation  $R$  with  $(s, t) \in R$ .



## Probabilistic bisimulation

### Probabilistic bisimulation

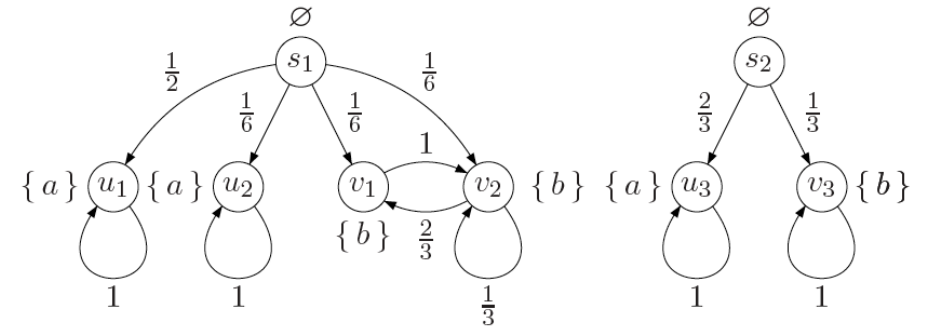
Let  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  be a DTMC and  $R \subseteq S \times S$  an **equivalence**. Then:  $R$  is a **probabilistic bisimulation** on  $S$  if for any  $(s, t) \in R$ :

1.  $L(s) = L(t)$ , and
2.  $\mathbf{P}(s, C) = \mathbf{P}(t, C)$  for all equivalence classes  $C \in S/R$ .

### Remarks

As opposed to bisimulation on states in transition systems, **any** probabilistic bisimulation is an equivalence.

## Example



## Quotient under $\sim_p$

### Quotient DTMC under $\sim_p$

For  $\mathcal{D} = (S, \mathbf{P}, \iota_{\text{init}}, AP, L)$  and probabilistic bisimulation  $\sim_p \subseteq S \times S$  let

$$\mathcal{D}/\sim_p = (S', \mathbf{P}', \iota'_{\text{init}}, AP, L'), \quad \text{the \textit{quotient} of } \mathcal{D} \text{ under } \sim_p$$

where

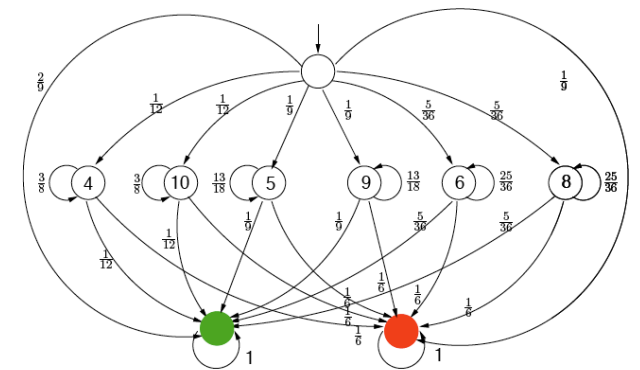
- ▶  $S' = S/\sim_p = \{[s]_{\sim_p} \mid s \in S\}$  with  $[s]_{\sim_p} = \{s' \in S \mid s \sim_p s'\}$
- ▶  $\mathbf{P}'([s]_{\sim_p}, [s']_{\sim_p}) = \mathbf{P}(s, [s']_{\sim_p})$
- ▶  $\iota'_{\text{init}}([s]_{\sim_p}) = \sum_{s' \in [s]_{\sim_p}} \iota_{\text{init}}(s')$
- ▶  $L'([s]_{\sim_p}) = L(s)$ .

### Remarks

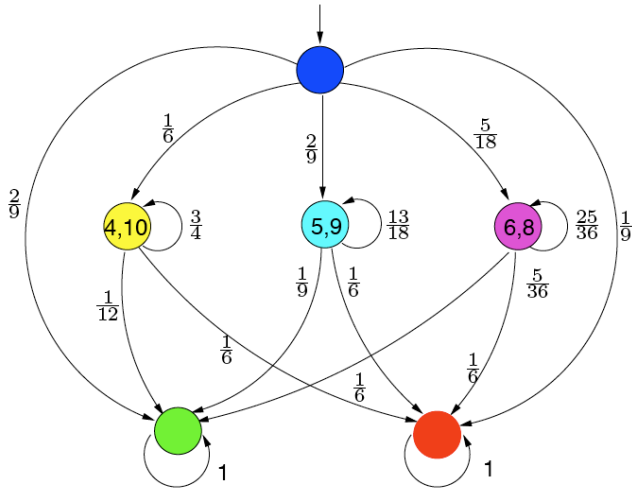
The transition probability from  $[s]_{\sim_p}$  to  $[t]_{\sim_p}$  equals  $\mathbf{P}(s, [t]_{\sim_p})$ . This is well-defined as  $\mathbf{P}(s, C) = \mathbf{P}(s', C)$  for all  $s \sim_p s'$  and all bisimulation equivalence classes  $C$ .

## Craps

- ▶ Come-out roll:
  - ▶ 7 or 11: win
  - ▶ 2, 3, or 12: lose
  - ▶ else: roll again
- ▶ Next roll(s):
  - ▶ 7: lose
  - ▶ point: win
  - ▶ else: roll again



## Quotient DTMC of Craps under $\sim_p$



## PCTL\* syntax

### Probabilistic Computation Tree Logic: Syntax

PCTL\* consists of state- and path-formulas.

- ▶ PCTL\* *state formulas* over the set  $AP$  obey the grammar:

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg \Phi \mid \mathbb{P}_J(\varphi)$$

where  $a \in AP$ ,  $\varphi$  is a path formula and  $J \subseteq [0, 1]$ ,  $J \neq \emptyset$  is a non-empty interval.

- ▶ PCTL\* *path formulae* are formed according to the following grammar:

$$\varphi ::= \Phi \mid \neg \varphi \mid \varphi_1 \wedge \varphi_2 \mid \bigcirc \varphi \mid \varphi_1 \mathbf{U} \varphi_2$$

where  $\Phi$  is a state formula and  $\varphi$ ,  $\varphi_1$ , and  $\varphi_2$  are path formulae.

## Preservation of PCTL-formulas

### Bisimulation preserves PCTL

Let  $\mathcal{D}$  be a DTMC and  $s, t$  states in  $\mathcal{D}$ . Then:

$$s \sim_p t \quad \text{if and only if} \quad s \text{ and } t \text{ are PCTL-equivalent.}$$

### Remarks

$s \sim_p t$  implies that

1. transient probabilities, reachability probabilities,
2. repeated reachability, persistence probabilities
3. all qualitative PCTL formulas

for  $s$  and  $t$  are equal.

If for PCTL-formula  $\Phi$  we have  $s \models \Phi$  but  $t \not\models \Phi$ , then it follows  $s \not\sim_p t$ . A *single* PCTL-formula suffices!

## Bounded until in PCTL\*

### Bounded until

Bounded until can be defined using the other operators:

$$\varphi_1 \mathbf{U}^{\leq n} \varphi_2 = \bigvee_{0 \leq i \leq n} \psi_i \quad \text{where } \psi_0 = \varphi_2 \text{ and } \psi_{i+1} = \varphi_1 \wedge \bigcirc \psi_i \text{ for } i \geq 0.$$

### Examples in PCTL\* but not in PCTL

$$\mathbb{P}_{> \frac{1}{4}}(\bigcirc a \mathbf{U} \bigcirc b) \text{ and } \mathbb{P}_{=1}(\mathbb{P}_{> \frac{1}{2}}(\Box \Diamond a) \vee \mathbb{P}_{\leq \frac{1}{3}}(\Diamond \Box b)).$$

## Preservation of PCTL\*-formulas

### Bisimulation preserves PCTL\*

Let  $\mathcal{D}$  be a DTMC and  $s, t$  states in  $\mathcal{D}$ . Then:

$s \sim_p t$  if and only if  $s$  and  $t$  are PCTL\*-equivalent.

### Remarks

1. Bisimulation thus preserves not only all PCTL but also all PCTL\* formulas.
2. By the last two results it follows that PCTL- and PCTL\*-equivalence coincide. Thus any two states that satisfy the same PCTL formulas, satisfy the same PCTL\* formulas.

## Preservation of PCTL

### PCTL/PCTL\* and Bisimulation Equivalence

Let  $\mathcal{D}$  be a DTMC and  $s_1, s_2$  states in  $\mathcal{D}$ . Then, the following statements are equivalent:

- (a)  $s_1 \sim_p s_2$ .
- (b)  $s_1$  and  $s_2$  are PCTL\*-equivalent, i.e., fulfill the same PCTL\* formulas
- (c)  $s_1$  and  $s_2$  are PCTL-equivalent, i.e., fulfill the same PCTL formulas
- (d)  $s_1$  and  $s_2$  are PCTL<sup>-</sup>-equivalent, i.e., fulfill the same PCTL<sup>-</sup> formulas

### Proof:

1. (a)  $\implies$  (b): by structural induction on PCTL\* formulas.
2. (b)  $\implies$  (c): trivial as PCTL is a sublogic of PCTL\*.
3. (c)  $\implies$  (d): trivial as PCTL<sup>-</sup> is a sublogic of PCTL.
4. (d)  $\implies$  (a): involved. First finite DTMCs, then for arbitrary DTMCs.

## PCTL<sup>-</sup> syntax

### Simple Probabilistic Computation Tree Logic: Syntax

PCTL<sup>-</sup> only consists of state-formulas. These formulas over the set  $AP$  obey the grammar:

$$\Phi ::= a \mid \Phi_1 \wedge \Phi_2 \mid \mathbb{P}_{\leq p}(\bigcirc \Phi)$$

where  $a \in AP$  and  $p$  is a probability in  $[0, 1]$ .

### Remarks

This is a truly simple logic. It does not contain the until-operator. Negation is **not** present and cannot be expressed. Only upper bounds on probabilities.

The next theorem shows that PCTL-, PCTL\*- and PCTL<sup>-</sup>-equivalence **coincide**.

## IEEE 802.11 group communication protocol

	original DTMC			quotient DTMC		red. factor	
OD	states	transitions	ver. time	blocks	total time	states	time
4	1125	5369	122	71	13	<b>15.9</b>	<b>9.00</b>
12	37349	236313	7180	1821	642	<b>20.5</b>	<b>11.2</b>
20	231525	1590329	50133	10627	5431	<b>21.8</b>	<b>9.2</b>
28	804837	5750873	195086	35961	24716	<b>22.4</b>	<b>7.9</b>
36	2076773	15187833	5103900	91391	77694	<b>22.7</b>	<b>6.6</b>
40	3101445	22871849	7725041	135752	127489	<b>22.9</b>	<b>6.1</b>

## Summary

- ▶ Bisimilar states have equal transition probabilities to all equivalence classes.
- ▶  $\sim_p$  is the coarsest probabilistic bisimulation.
- ▶ In a quotient DTMC all states are equivalence classes under  $\sim_p$ .
- ▶ Bisimulation, i.e.,  $\sim_p$ , and PCTL-equivalence coincide.
- ▶ PCTL, PCTL\* and PCTL<sup>-</sup>-equivalence coincide.
- ▶ To show  $s \not\sim_p t$ , show  $s \models \Phi$  and  $t \not\models \Phi$  for  $\Phi \in \text{PCTL}^-$ .
- ▶ Bisimulation may yield up to exponential savings in state space.

### Take-home message

Probabilistic bisimulation coincides with a notion from the sixties, named (ordinary) lumpability.