

# Reachability Probabilities in Markov Chains

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# Overview

- 1 Motivation
- 2 What are discrete-time Markov chains?
- 3 Reachability probabilities



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- ▶ When analysing system performance and dependability
  - ▶ to quantify arrivals, waiting times, time between failure, QoS, ...



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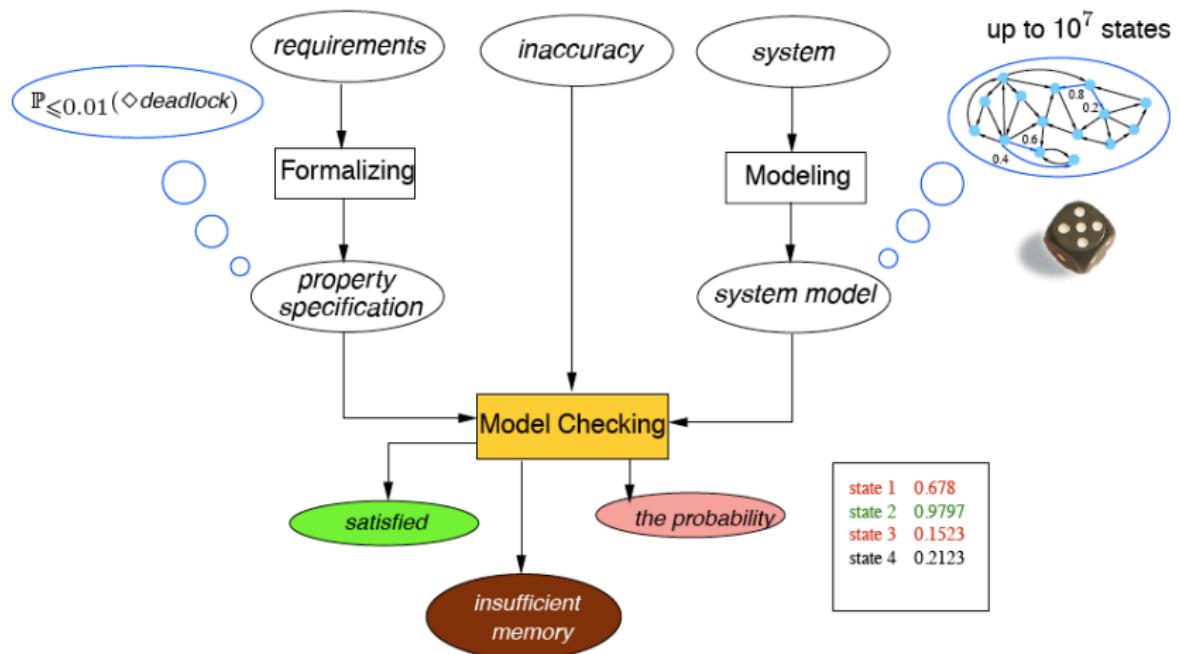
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- ▶ When building protocols for networked embedded systems
  - ▶ randomized algorithms
- ▶ When problems are undecidable deterministically
  - ▶ repeated reachability of lossy channel systems, ...

# What is probabilistic model checking?



# 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)



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# DTMCs — A transition system perspective

## Discrete-time Markov chain

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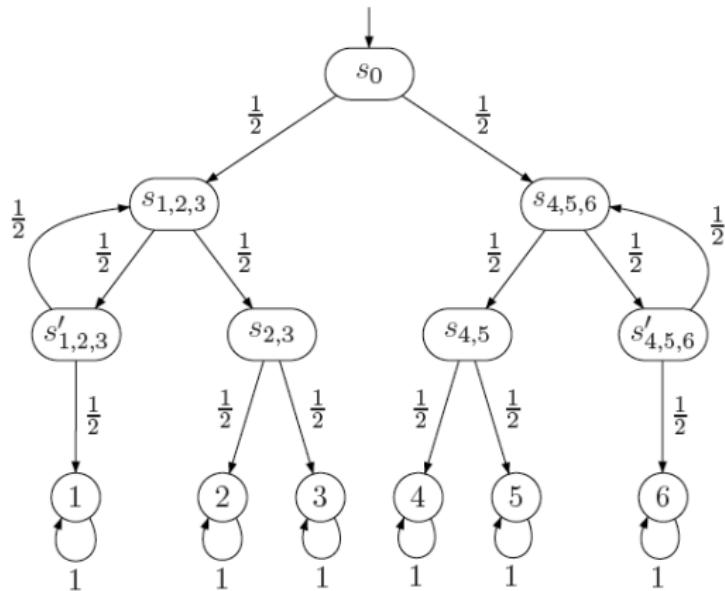
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## Initial states

- ▶  $\iota_{\text{init}}(s)$  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

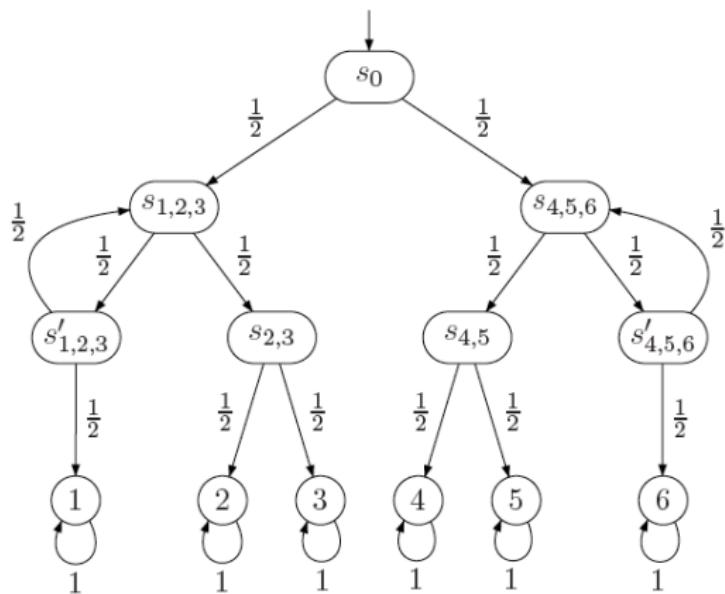
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A geometric distribution is the *only* discrete probability distribution that is memoryless.

# Determining $n$ -step transition probabilities

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

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

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

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

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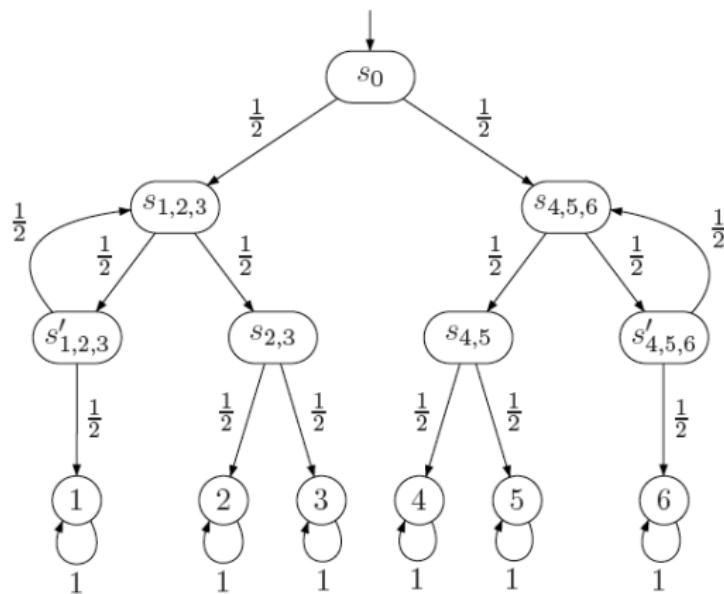
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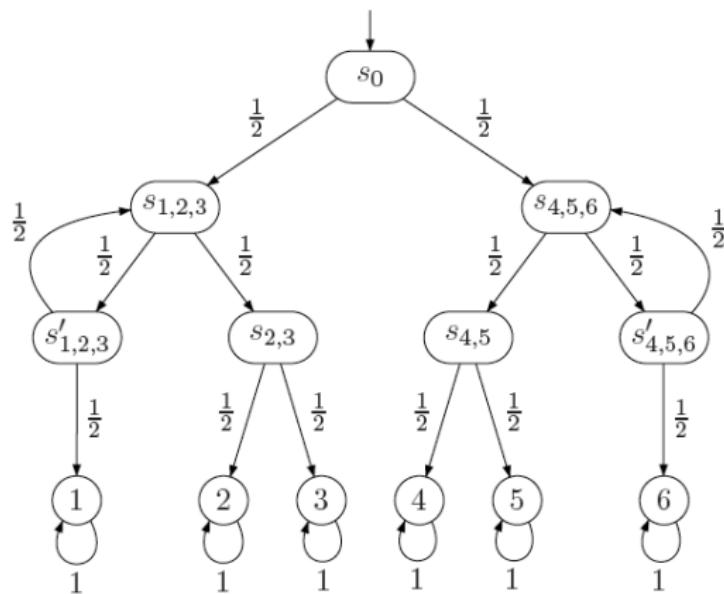
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3. ... and they are transient probabilities in a slightly modified DTMC.

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# Paths

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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  $\mathbf{P}(s, s') > 0$ .

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

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## Repeated reachability

Repeatedly visit a state in  $G$ ; formally:

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## Repeated reachability

Repeatedly visit a state in  $G$ ; formally:

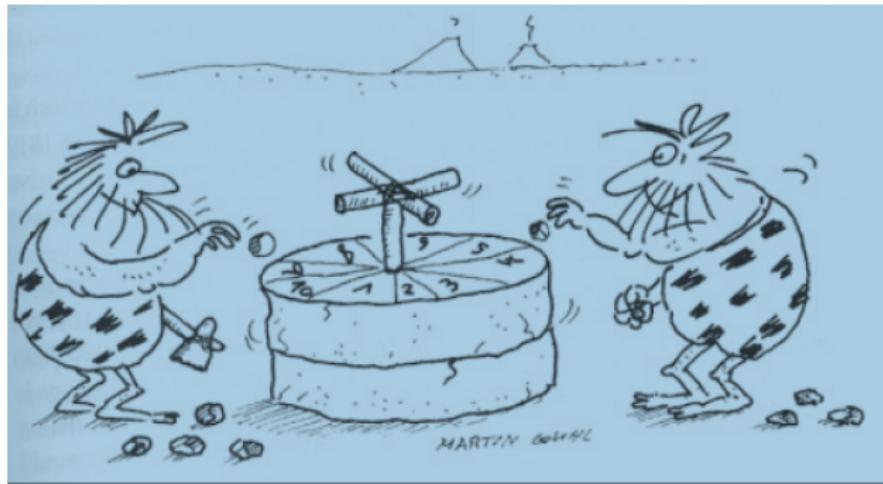
$$\square \diamond G = \{ \pi \in \text{Paths}(\mathcal{D}) \mid \forall i \in \mathbb{N}. \exists j \geq i. \pi[j] \in G \}$$

## Persistence

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

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# What's the probability of infinite paths?



# Paths and probabilities

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- ▶ Cylinder set of a finite path := set of all its infinite continuations.

# Probability measure on DTMCs

## Cylinder set

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

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## Probability space of a DTMC

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

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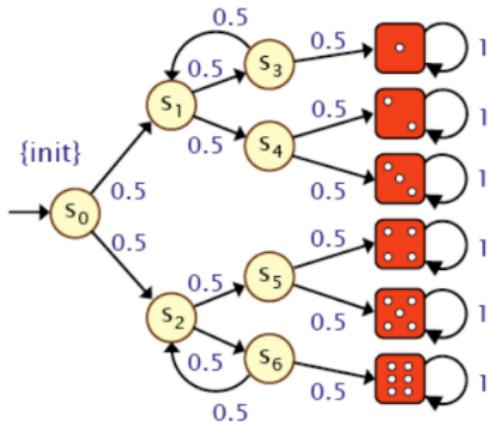
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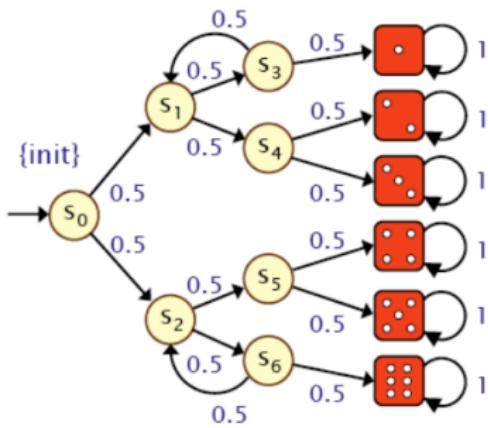
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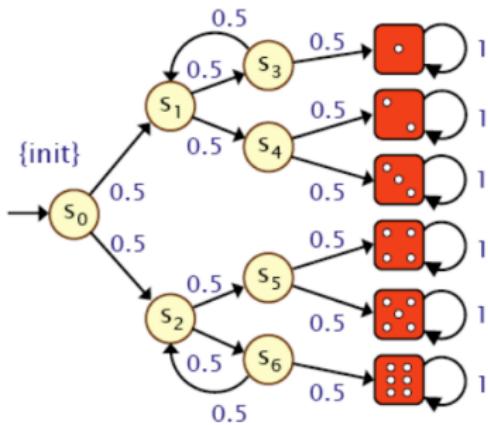
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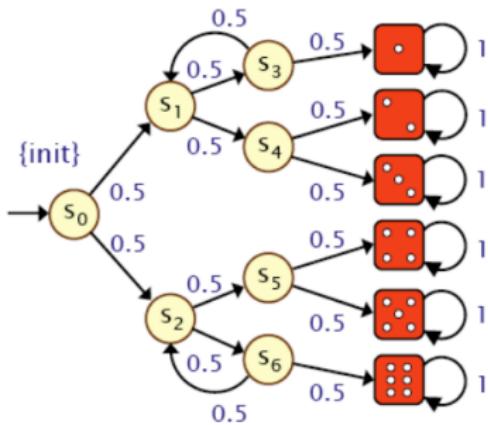


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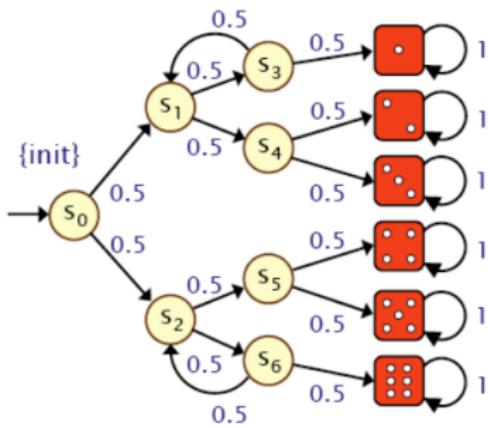
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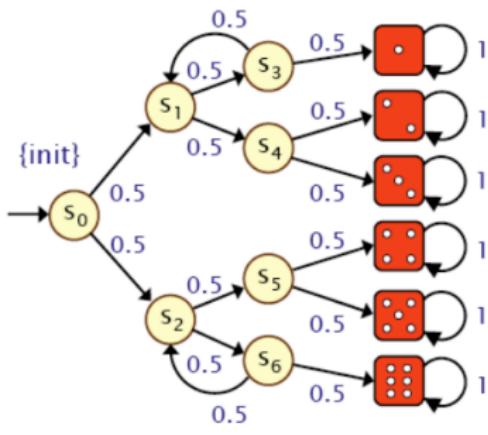


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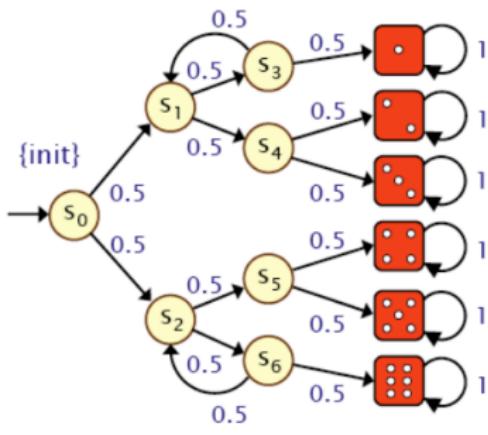


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There is however an **simpler** way to obtain reachability probabilities!

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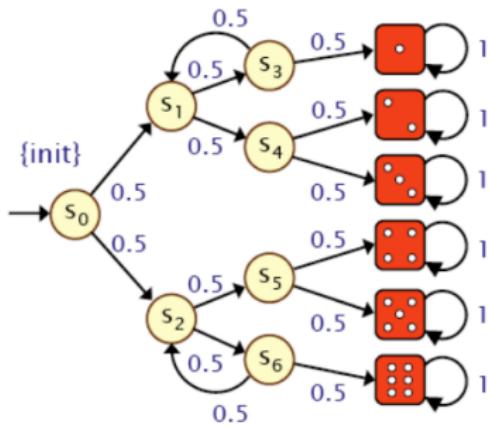
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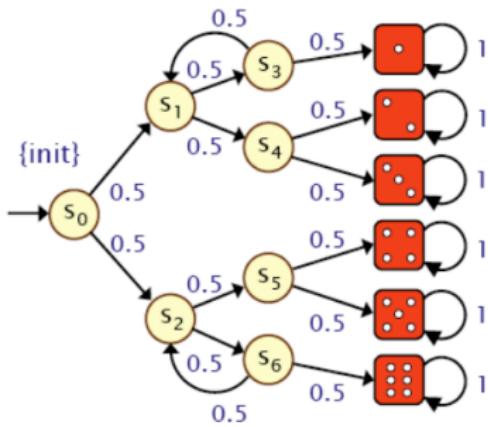
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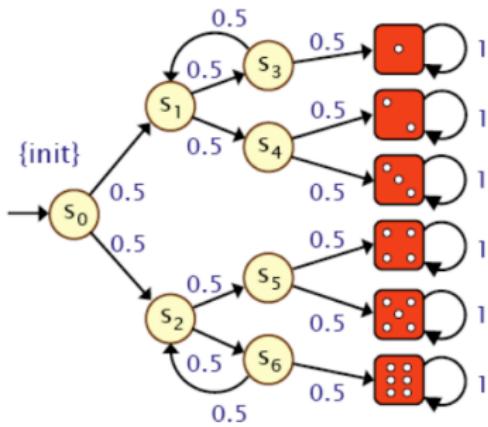


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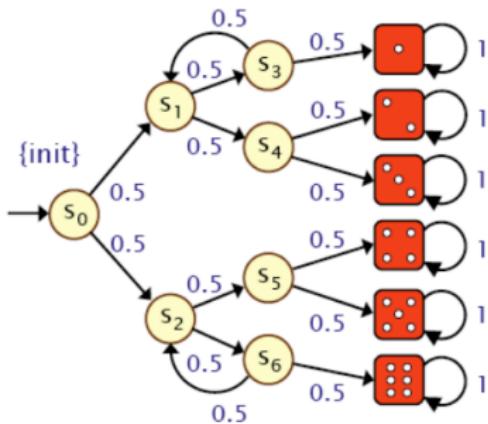
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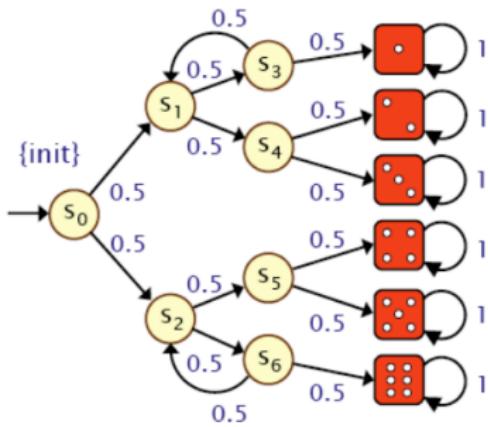
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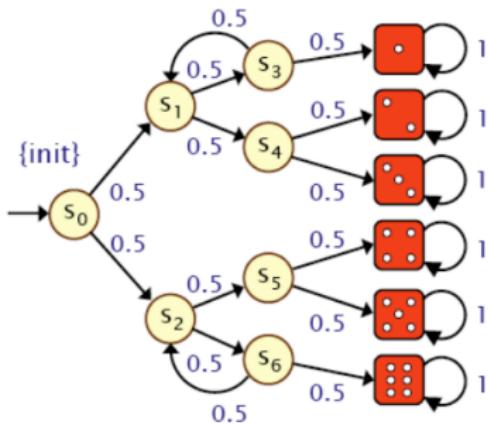
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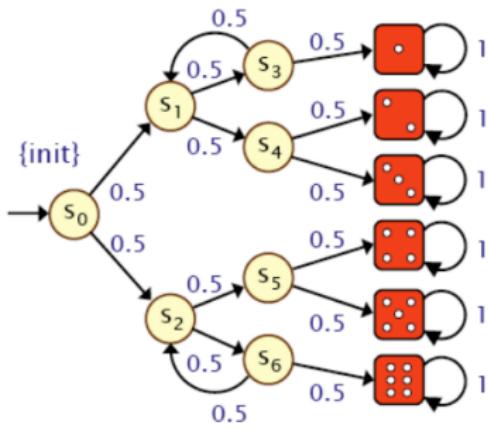
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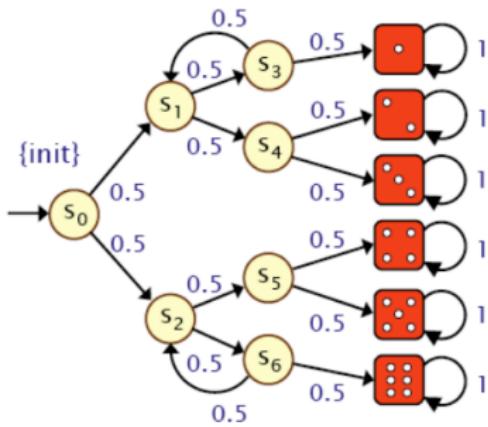
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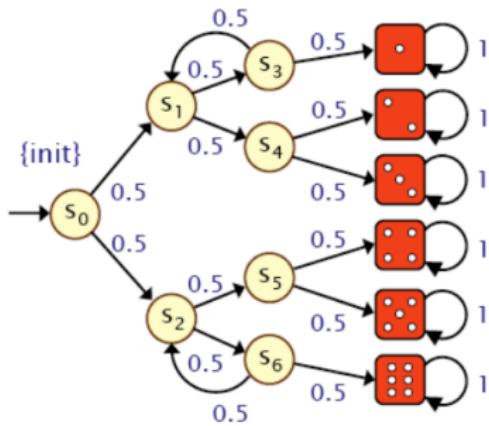
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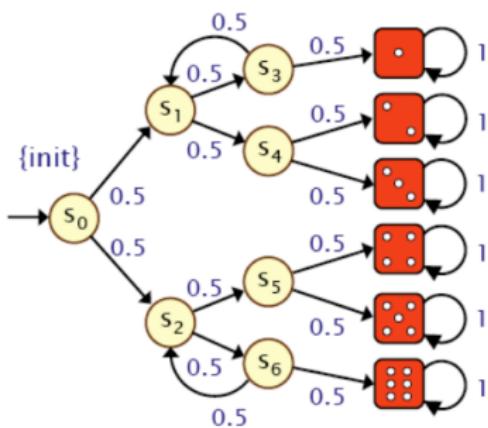
where  $\mathbf{I}$  is the identity matrix of cardinality  $|S_?| \times |S_?|$ .

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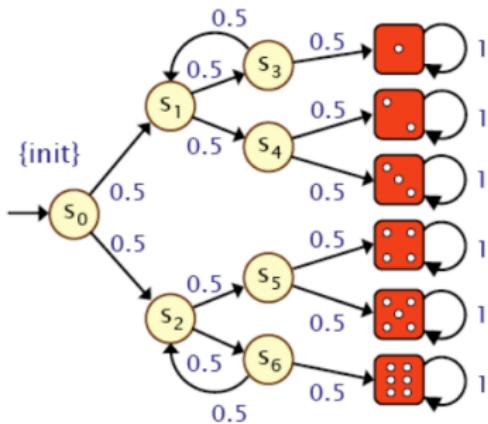


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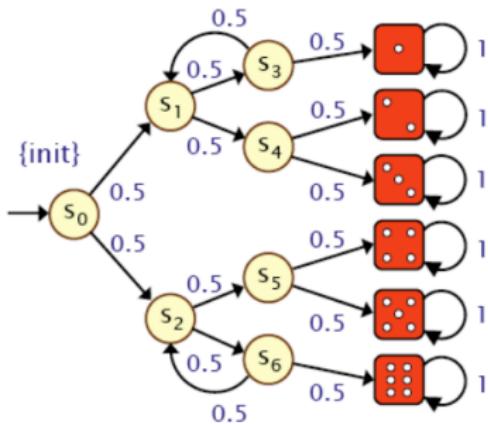
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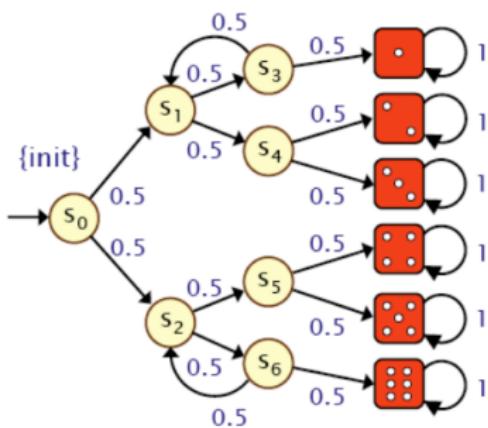
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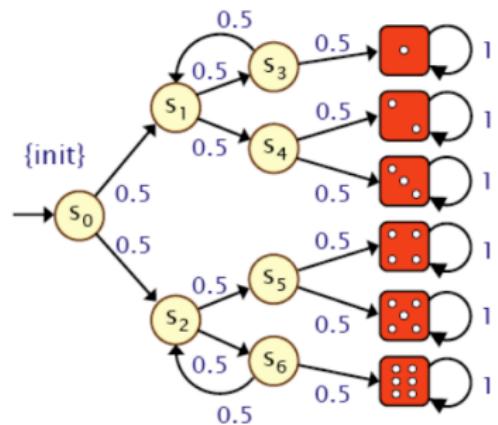
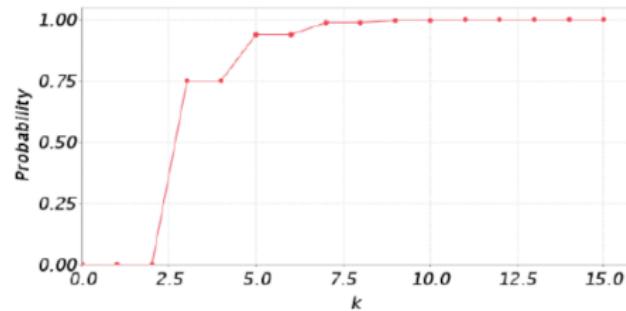
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Alternatives: e.g., Jacobi or Gauss-Seidel, successive overrelaxation (SOR).

# 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:



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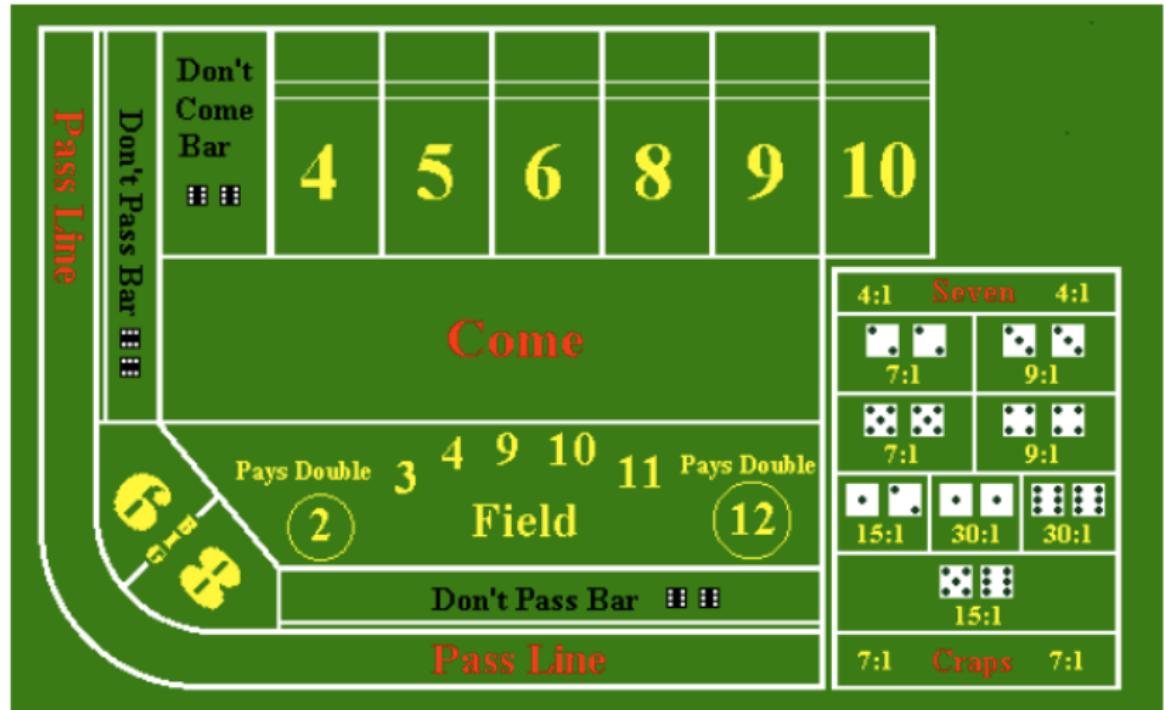
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$$\underbrace{\Pr(\overline{F} \cup^{\leq n} G)}_{\text{reachability in } \mathcal{D}} = \underbrace{\Pr(\Diamond^{\leq n} G)}_{\text{reachability in } \mathcal{D}[F \cup G]} = \underbrace{\iota_{\text{init}} \cdot \mathbf{P}_{F \cup G}^n}_{\text{in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

# Spare time tonight? Play Craps!



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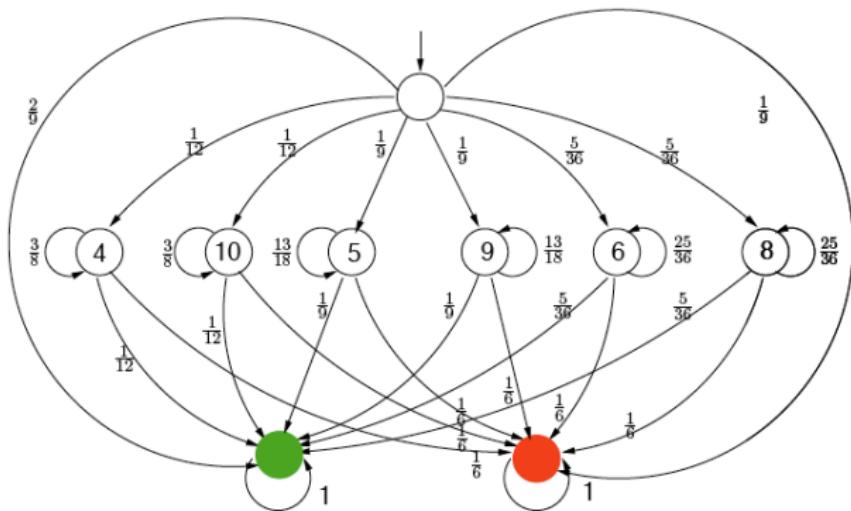


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What is the probability to win the Craps game?

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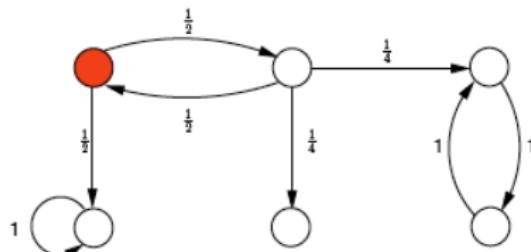
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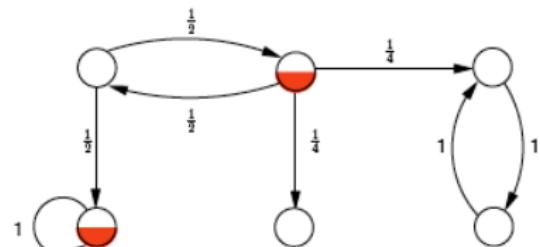
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4. ... and they are **transient probabilities** in a slightly modified DTMC.

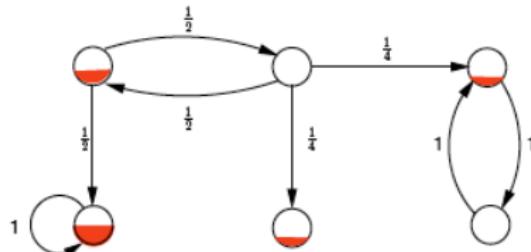
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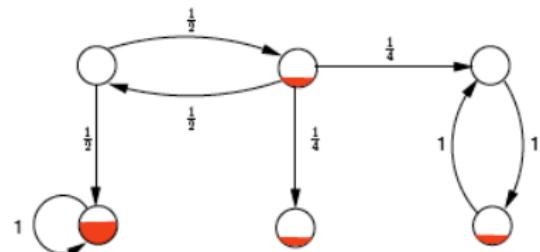
zero-th epoch



first epoch

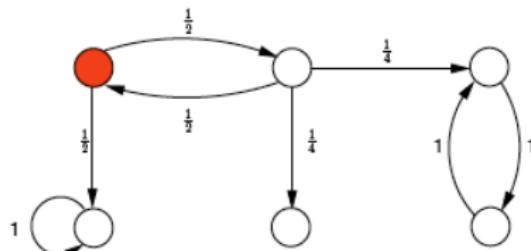


second epoch

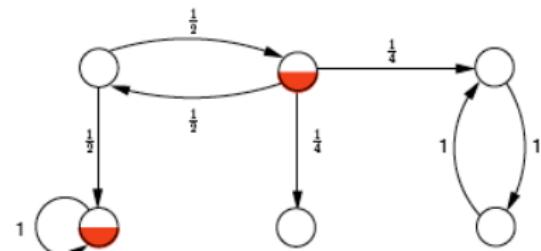


third epoch

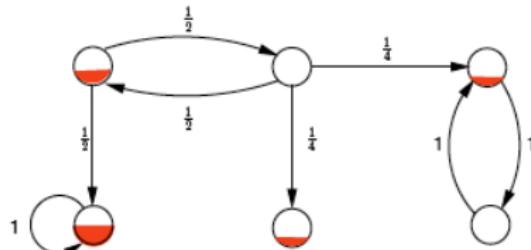
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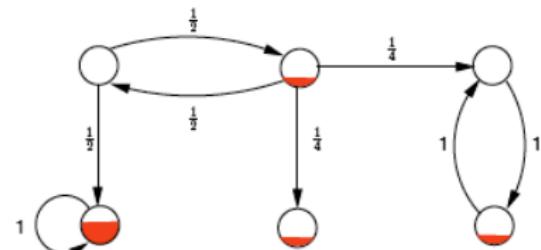
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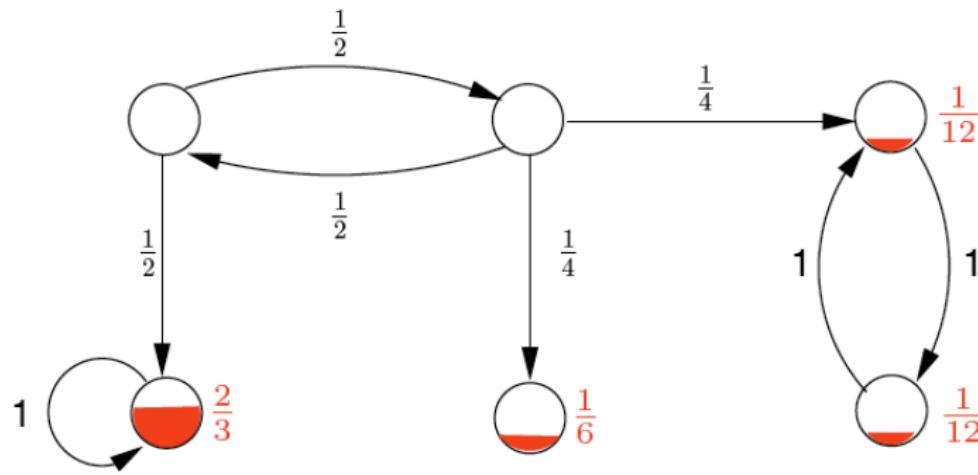
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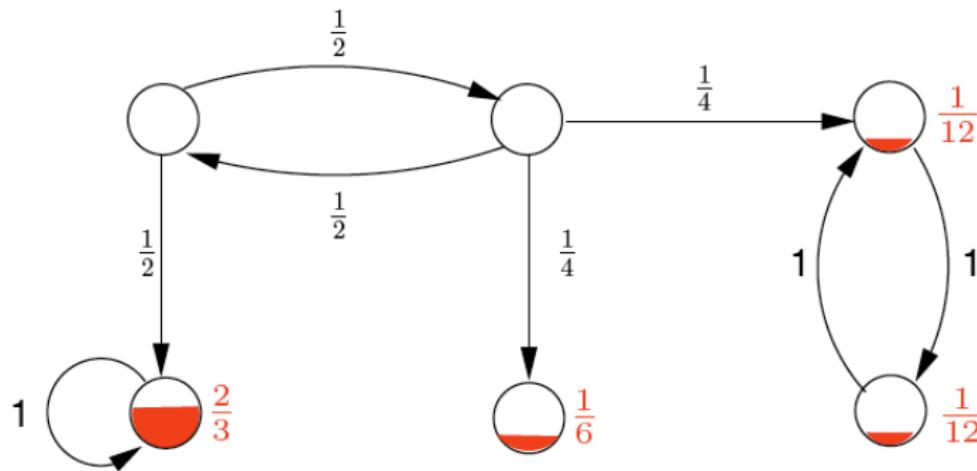
third epoch

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

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

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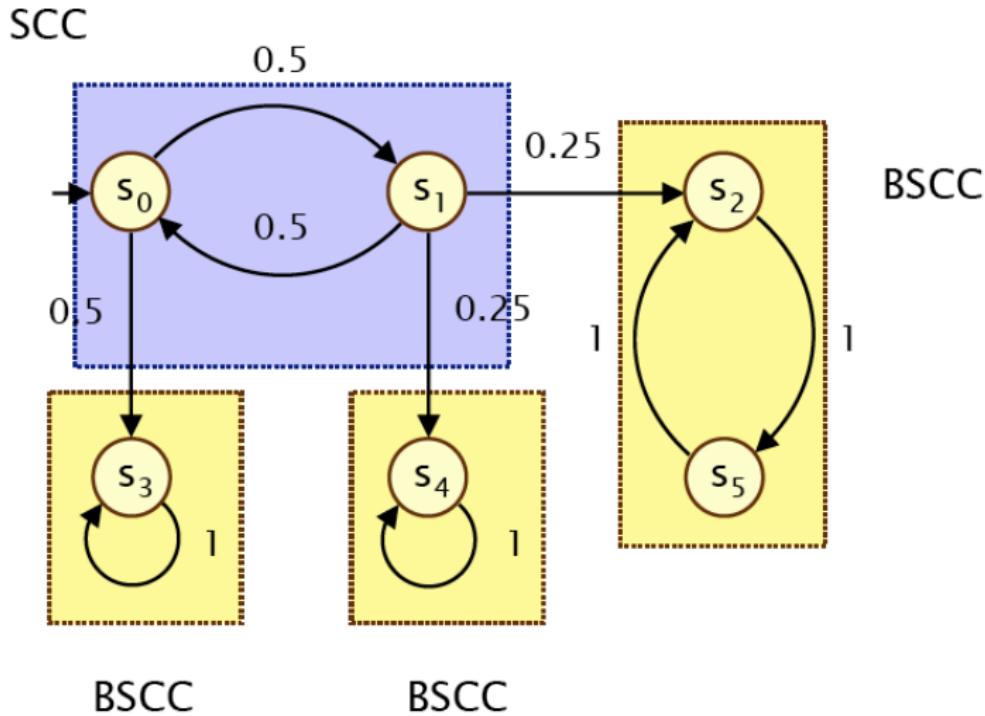
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- ▶ SCC  $T$  is a *bottom SCC* (BSCC) if no state outside  $T$  is reachable from  $T$ , i.e., for any state  $s \in T$ ,  $\mathbf{P}(s, T) = \sum_{t \in T} \mathbf{P}(s, t) = 1$ .

# Example



# Long-run theorem

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

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

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## Remark

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

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

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Consider a finite Markov chain  $\mathcal{D}$  with state space  $S$ ,  $\textcolor{blue}{G} \subseteq S$ , and  $s \in S$ .

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