

Model checking PCTL on MDPs

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Let's play a better game



Recapitulation

- **Markov decision processes** combine nondeterministic and probabilistic choices.
- **Strategies** select how to resolve nondeterministic choices.
- MDP + strategy **induce** a Markov chain.
- Probabilities on a MDP are defined via the induced Markov chain.

Formal definition

- A Markov decision process consists of:
 - S finite set of states
 (often $S = \{1, 2, \dots, n\}$)
 - Act finite set of actions
 - $\mathbf{P}: S \times Act \times S \rightarrow [0,1]$
 transition probability matrix
 for every action
 - $L: S \rightarrow 2^{AP}$ labelling with atomic propositions

Probabilistic CTL

- a logic to describe properties of Markov chains and MDPs
- extends CTL
- strictly speaking, not a quantitative logic (truth values are Boolean: true or false)

PCTL syntax for MDPs

- state formulas ϕ, ψ
 - a atomic proposition
 - $\neg\phi$ negation
 - $\phi \vee \psi$ disjunction
 - $\mathbf{P}_{\leq p}(\text{E}\Pi), \mathbf{P}_{\geq p}(\text{E}\Pi), \mathbf{P}_{\leq p}(\text{A}\Pi), \mathbf{P}_{\geq p}(\text{A}\Pi), \dots$
probabilistic operator
- path formulas Π
 - $X\phi$ next state
 - $\phi U\psi$ unbounded until
 - $\phi U^{\leq k}\psi$ bounded until

What strategy to choose?

- the best strategy: highest probability to win
- the worst strategy: lowest probability to win

- add operators A, E similar to CTL
 - A = for all strategies
 - E = for some strategy

- often simplified to: implicitly A only

Model checking PCTL for MDPs

- find strategy and probabilities at the same time
- three methods:
 - value iteration (similar to \mathcal{V})
 - linear programming (inequality system)
 - policy iteration
(pick a random action and then improve)

What strategy is needed?

- history-dependent strategy? simpler strategy?
- unbounded until: history-independent
- bounded until: step-dependent
- Proofs on the following slides...

Today's programme

- algorithm for model checking
 - general: bottom-up
 - bounded until
(needs a step-dependent scheduler)
 - unbounded until (history-independent scheduler)
- examples

Model checking PCTL formulas

set of states
that satisfy ϕ

Calculate $\text{Sat}(\phi)$ from subformulas:

- $\text{Sat}(a) = \{s \in S \mid a \in L(s)\}$
- $\text{Sat}(\neg\varphi) = S \setminus \text{Sat}(\varphi)$
- $\text{Sat}(\varphi \vee \psi) = \text{Sat}(\varphi) \cup \text{Sat}(\psi)$
- $\text{Sat}(\mathbf{P}_{\geq p}(\mathbf{A}\Pi)) = \{s \mid \inf_{\eta \in \text{Estrat}} \text{Prob}_s^\eta \{\sigma \mid \sigma \models \Pi\} \geq p\}$
- $\text{Sat}(\mathbf{P}_{\geq p}(\mathbf{E}\Pi)) = \{s \mid \sup_{\eta \in \text{Estrat}} \text{Prob}_s^\eta \{\sigma \mid \sigma \models \Pi\} \geq p\}$
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notatie voor schedulers...

- griekse letter, bv. η
- Frakturletter, bv. \mathfrak{A} , \mathfrak{S}



Path formula: X (next)

$$\sup_{\eta \in \text{Strat}} \text{Prob}_s^\eta \{ \sigma \mid \sigma \models X\phi \}$$

$$= \max_{\alpha \in \text{Act}} \mathbf{P}(s, \alpha, \text{Sat}(\phi))$$

$$= \max_{\alpha \in \text{Act}} \sum_{t \in \text{Sat}(\phi)} \mathbf{P}(s, \alpha, t)$$

Path formula: $U^{\leq n}$ (bounded until)

- $\varphi U^{\leq k} \psi$
- Find
 - $S_1 =$ winning states 
 - $S_0 =$ losing states (may add stalemate states) 
 - $S_?$ = other states (including announce checkmate)
- Which strategy to choose?

Step-bounded strategies

- A step-bounded strategy bases its decision on:
 - current state
 - number of steps taken until now
 - no more information on past states!
- Theorem: **Step-bounded strategies suffice to find the best strategy for bounded until.**
(Proof idea: number of steps that remain before deadline is the only relevant data.)

Proof:

step-bounded strategies suffice

- proof by induction on number of remaining steps
 - base case: 0 steps remain.
No transition allowed \rightarrow class of scheduler of no influence
 - induction step: Assume given strategies η_s for each state s that are optimal for $\varphi U^{\leq n} \psi$.
To find an optimal strategy from s_0 for $\varphi U^{\leq n+1} \psi$, choose an action that leads to the highest probability under η_s .

Bounded reachability

- Assume $S_1 = \text{Sat}(\psi)$ (no announce checkmate)
- $x_n(s) :=$ maximal probability of $(\varphi U^{\leq n} \psi)$,
if s is the start state

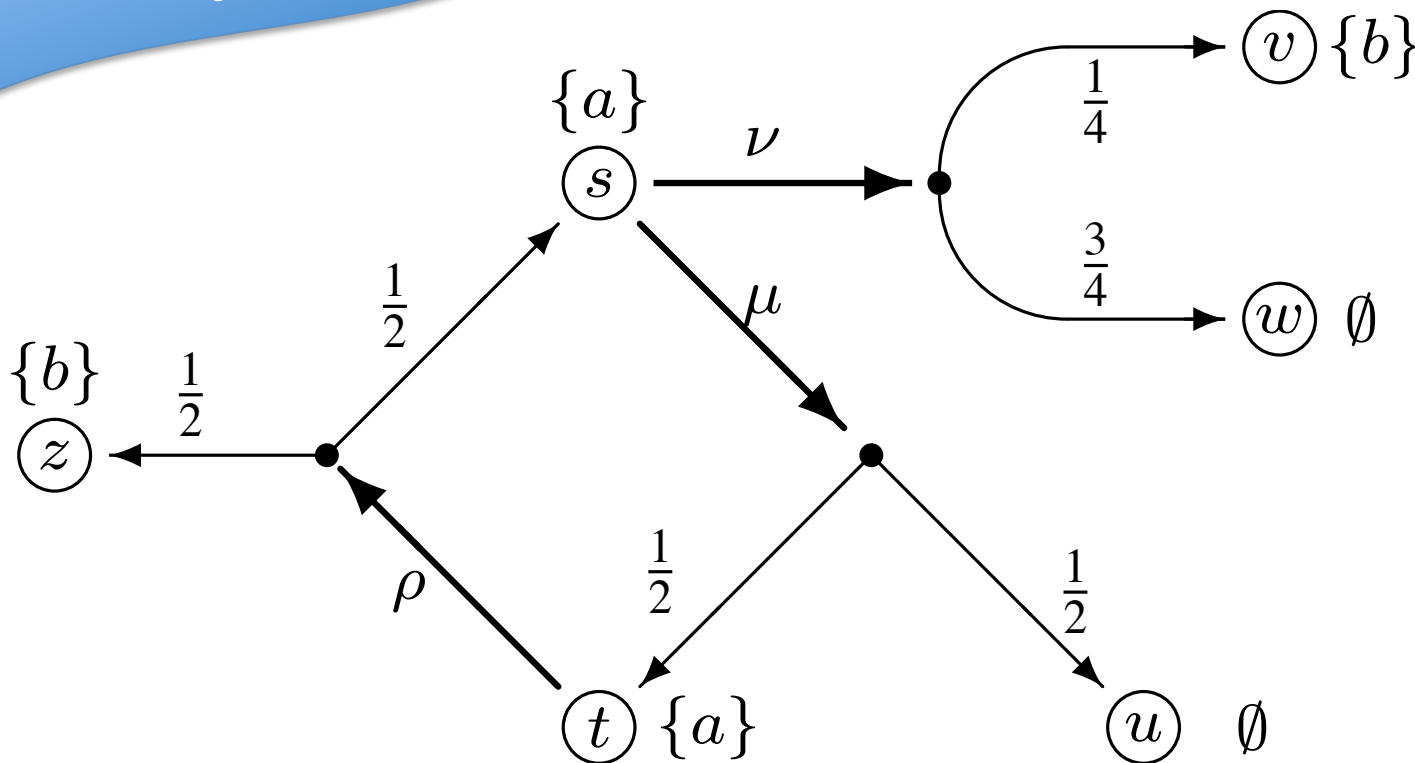
- $x_0(S_1) = 1$ $x_0(S_0 \cup S_?) = 0$

- $x_{n+1}(S_1) = 1$ $x_{n+1}(S_0) = 0$

$$x_{n+1}(s) = \max_{\alpha \in \text{Act}} \sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_n(t) \quad \text{for } s \in S_?$$

The example also shows that simple strategies do not suffice!

Example



Which states satisfy $\mathbf{P}_{>1/4}(a \text{ EU}^{\leq 3} b)$?

Path formula: U (unbounded until)

- $\varphi U \psi$
- Find
 - S_1 = winning and announce checkmate states
 - S_0 = losing and stalemate states
 - $S_?$ = other states
- Which strategy to choose?

An equation system...

- $x(s) :=$ maximal probability of $(\varphi \cup \psi)$,
if s is the start state, under a *simple* strategy
- $x(S_1) = 1$ $x(S_0) = 0$

$$x(s) = \max_{\alpha \in Act} \sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x(t) \quad \text{for } s \in S_?$$

...but what about other strategies?

- Assume
 $y(s) :=$ maximal probability of $(\varphi \cup \psi)$,
if s is the start state, under any strategy
- $y(s)$ must be a solution of this equation system!
- uniqueness of solution $\rightarrow x = y$

Value iteration + convergence test

- $x_n(s) :=$ lower bound on maximal prob of $(\varphi \cup \psi)$,
 $x^n(s) :=$ upper bound on maximal prob of $(\varphi \cup \psi)$,
if s is the start state
- $x_0(S_1) = x^0(S_1 \cup S_?) = 1, \quad x_0(S_0 \cup S_?) = x^0(S_0) = 0$
- $x_{n+1}(S_1) = x^{n+1}(S_1) = 1, \quad x_{n+1}(S_0) = x^{n+1}(S_0) = 0$

$$x_{n+1}(s) = \max_{\alpha \in Act} \sum_{t \in S} P(s, \alpha, t) \cdot x_n(t) \quad \text{for } s \in S_?$$

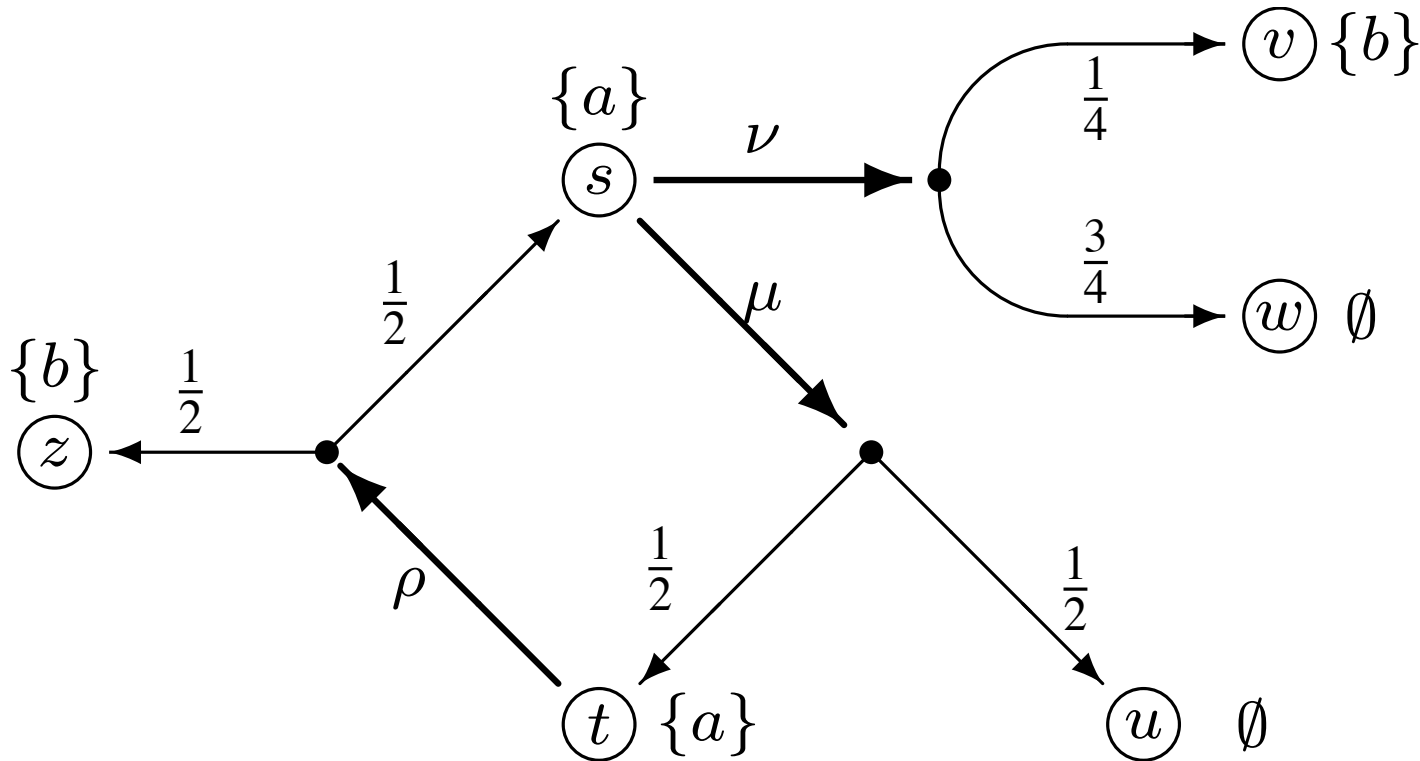
$$x^{n+1}(s) = \max_{\alpha \in Act} \sum_{t \in S} P(s, \alpha, t) \cdot x^n(t) \quad \text{for } s \in S_?$$

Policy iteration

- Start with a random strategy ζ_0
- Calculate $x_{\zeta_0}(s) := Prob^{\zeta_0}_s(\varphi \cup \psi)$
- For each state, find the optimal action if the other states use ζ_n :

$$\zeta_{n+1}(s) = \arg \max_{\alpha \in Act} \sum_{t \in S} \mathbf{P}(s, \alpha, t) \cdot x_{\zeta_n}(t)$$

Example



Which states satisfy $\mathbf{P}_{\leq \frac{1}{2}}(a \text{ AU } b)$?

Some slides have been corrected
based on remarks made by Remy Viehoff.