

Probabilistic Model Checking

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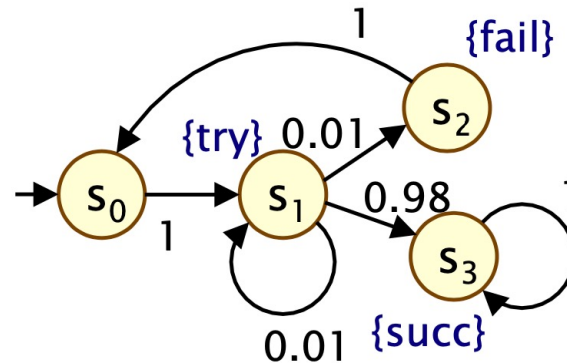
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additional slides DTMC

Probability of taking a path or a set of paths

A “DTMC” :

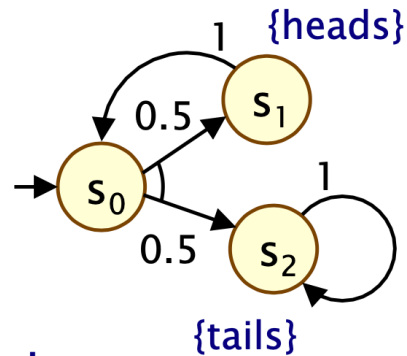


- Consider a path ω e.g. s_0, s_1, s_2, s_0 . The probability that the system follows this path when executed with the starting state s_0 is denoted by $P_{s_0}(\omega)$. Or simply $P(\omega)$ if it is clear which s_0 is meant. It is the product of the probability of each transition in ω .

Example: for the above ω , $P(\omega) = 1 * 0.01 * 1 = 0.01$

- For a **set of paths** U (starting from s_0), the probability that the system's execution follows **one of** the paths in U , denoted by $P(U)$, is $\sum_{\omega \in U} P(\omega)$.

Probability of taking a path or a set of paths

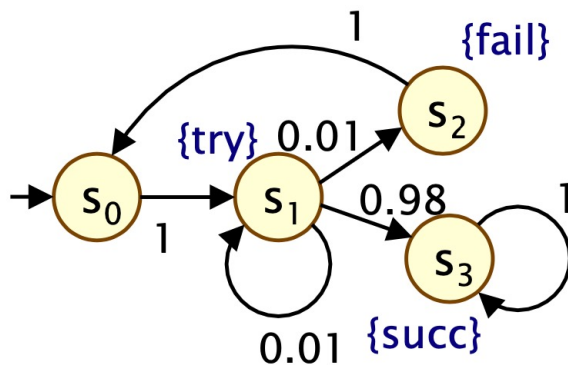


Example: consider U = the set of paths that ends in s_2 . Note that U is infinite: $U = \{ 02, 0102, 010102, \dots \}$. But we can calculate $P(U)$.

$$P(U) = 0.5 + 0.5^2 + 0.5^3 + \dots = \sum_{k \geq 0} 0.5^k$$

$$= 0.5 * \frac{1}{1-0.5} = 1$$

Probability Matrix Representation



P:

	s0	s1	s2	s3
s0	0	1	0	0
s1	0	0.01	0.01	0.98
s2	1	0	0	0
s3	0	0	0	1

$P_{i,k}$ = the value at the i -th row and k -th column. It specifies the probability of taking the transition $s_i \rightarrow s_k$, if we are now at s_i .

For example the circle red value above is $P_{1,2}$, specifying the probability of taking the transition from s_1 to s_2 (check the picture), which is 0.01.

Basic Operations on Probability Matrix

- Multiplying P with itself: P^n
- Multiplying a vector with P : $u \times P$
- Multiplying P with a vector: $P \times v$

Pⁿ

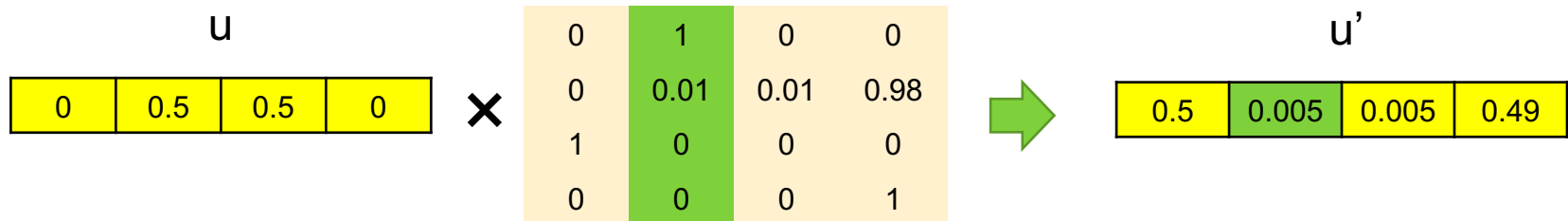
- $P^0 = I$ (identity matrix)
 $P^{n+1} = P \times P^n$
- $P^n_{i,k}$ is the probability of ending up in state s_k in n -steps, given we start in the state s_i .
- For example, with the previous P , let's look at P^2 :

0	1	0	0		0	1	0	0		?	?	0.01	?
0	0.01	0.01	0.98		0	0.01	0.01	0.98		?	?	?	?
1	0	0	0		1	0	0	0		?	?	?	?
0	0	0	1		0	0	0	1		?	?	?	?

$$\begin{aligned}
 P^2_{0,2} &= (P \times P)_{0,2} \\
 &= P_{0,0} * P_{0,2} + P_{0,1} * P_{1,2} + P_{0,2} * P_{2,2} + P_{0,3} * P_{3,2}
 \end{aligned}$$

Probability distribution of the next state, given the current distribution

- A **probability distribution** of the current state is the probability of currently being in various states. It can be given by a vector of size K , if K is the number of possible states. E.g. if $u = [0, 0.5, 0.5, 0]$ is the probability distribution of the current state, it says e.g. that there is 0.5 probability that currently we are in the state s_1 , but 0 probability that we are in the state s_0 .
- The product $u \times P$ (we often simply write it as uP) gives a new vector u' of size K , that gives us the probability distribution of the next state.



e.g. $u'_1 = u \cdot \text{the green column (dot product)}$

$$= u_0 * P_{0,1} + u_1 * P_{1,1} + u_2 * P_{2,1} + u_3 * P_{3,1}$$

Probability vector

- Sometimes we also want to know what the probability to end up in state, say, s_1 or s_2 as the **next** state, if we start in the state s_1 .
- We can represent “end up in either s_1 or s_2 ” with a vector $v = [0, 1, 1, 0]$.
- Let v^t is the *transpose* of v . The product $P \times v^t$ gives a w such that w is a (transposed) vector, where w_i is the probability to end up in one of the states specified in v , if we start in s_i .

0	1	0	0
0	0.01	0.01	0.98
1	0	0	0
0	0	0	1

 \times

0
1
1
0

 \rightarrow

1
0.02
0
0

v^t w

e.g. w_1 = the green row $\cdot v^t$ (dot product)
 $= P_{1,0} * v_0 + P_{1,1} * v_1 + P_{1,2} * v_2 + P_{1,3} * v_3$

Probability vector

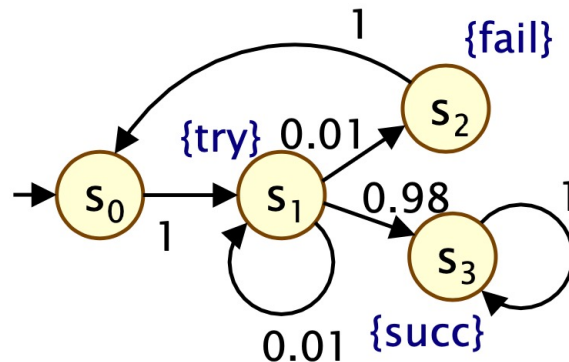
- Prob(φ) (notice the underscore) is a probability vector e.g. $w =$

1
0.02
0
0

such that the i -th element tells us what the probability that the system would behave as φ if executed in state s_i .

- Example: the above w (blue) happens to be equal to Prob($\mathbf{X}(\text{try } v \text{ fail})$).
- This notation Prob will be used later when we discuss model checking of probabilistic-CTL.

The construction of P' for bounded Until

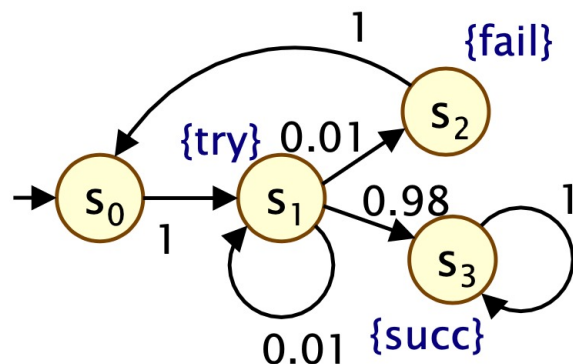


0	1	0	0
0	0.01	0.01	0.98
1	0	0	0
0	0	0	1

The probability matrix P of the DTMC on the left.

- Consider as an example to check whether the DTMC satisfies $P_{>0.99}[\text{try} \vee \neg \text{fail} \ U^{\leq 2} \text{succ}]$.
 - Calculate first the probability vector **Prob** $[\text{try} \vee \neg \text{fail} \ U^{\leq 2} \text{succ}]$.
 - From there you can calculate the set **Sat** $(P_{>0.99}[\text{try} \vee \neg \text{fail} \ U^{\leq 2} \text{succ}])$.
 - If the initial state s_0 is in the blue Sat-set then the property $P_{>0.99}[\text{try} \vee \neg \text{fail} \ U^{\leq 2} \text{succ}]$ holds on the DTMC.**

The construction of P' for bounded Until

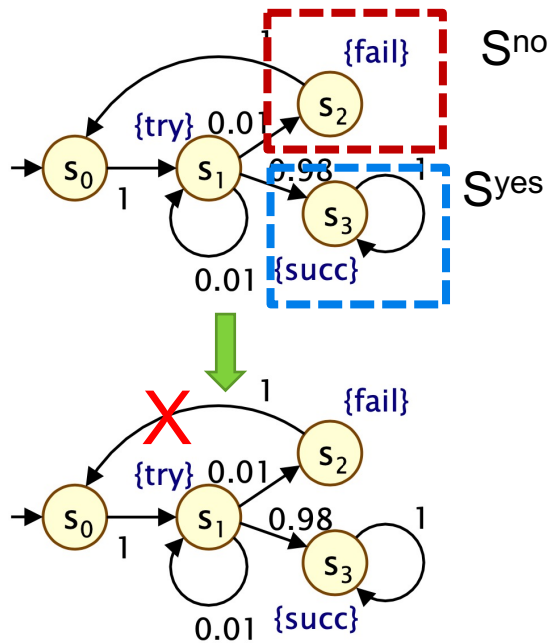


0	1	0	0
0	0.01	0.01	0.98
1	0	0	0
0	0	0	1

The probability matrix P of the DTMC on the left.

- To calculate the probability vector $\text{Prob}[\text{try} \vee \neg\text{fail} \ U^{\leq 2} \text{succ}]$, we would like to use the matrix P above, however it will also "contain" transitions that cause you to break the green-property. So the idea is to use a "modified" matrix P' .
- We pre-calculate first the $S^{\text{yes}} = \text{Sat}(\text{succ}) = \{s3\}$. On all states in S^{yes} , you have the green property immediately (in 0 step).
- We pre-calculate S^{no} , we take $S^{\text{no}} = \text{Sat}(\neg(\text{try} \vee \neg\text{fail}) \wedge \neg \text{succ}) = \{s2\}$. Executions starting from S^{no} won't satisfy your green-property above,

The construction of P' for bounded Until



$P :$

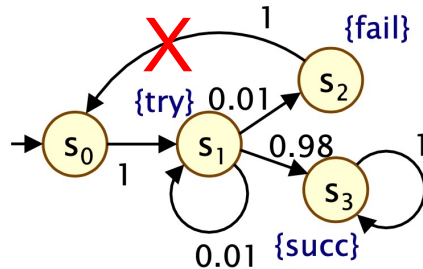
0	1	0	0
0	0.01	0.01	0.98
1	0	0	0
0	0	0	1

$P' :$

0	1	0	0
0	0.01	0.01	0.98
0	0	0	0
0	0	0	1

1. We remove outgoing arrows from the states in S^{no} and S^{yes} .
2. We keep all arrows that go out from states which are **not** in S^{no} nor S^{yes} .
3. We add a self-loop $s \rightarrow s$ with probability 1 for any state s in S^{yes} .

Using P' for bounded Until



P' :

0	1	0	0
0	0.01	0.01	0.98
0	0	0	0
0	0	0	1

- We now use P' to iteratively calculate $\text{Prob}[\text{try } \vee \neg\text{fail } \mathbf{U}^{\leq 2} \text{succ}]$

- From S^{yes} you know that $\text{Prob}[\text{try } \vee \neg\text{fail } \mathbf{U}^{\leq 0} \text{succ}] =$

0
0
0
1

- $\text{Prob}[\text{try } \vee \neg\text{fail } \mathbf{U}^{\leq 1} \text{succ}] = P' \times$

0
0
0
1

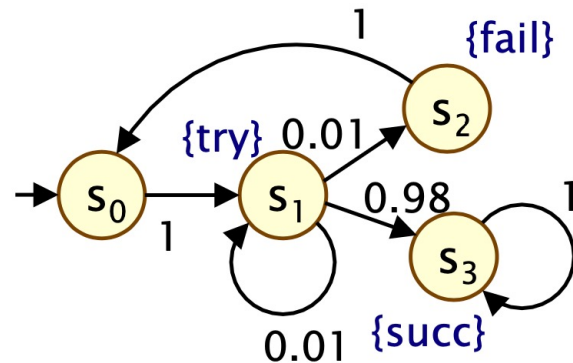
0
0.98
0
1

- $\text{Prob}[\text{try } \vee \neg\text{fail } \mathbf{U}^{\leq 2} \text{succ}] = P' \times$

0
0.98
0
1

0.98
0.9898
0
1

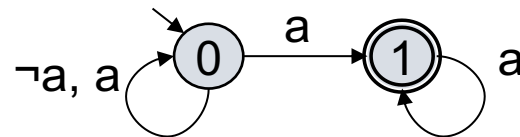
So, does the property hold?



- We have calculated $\text{Prob}[\text{try} \vee \neg\text{fail} \text{ } \mathbf{U}^{\leq 2} \text{ succ}] = [0.98, 0.9898, 0, 1]$
- So, the set $\text{Sat}(P_{>0.99}[\text{try} \vee \neg\text{fail} \text{ } \mathbf{U}^{\leq 2} \text{ succ}]) = \{s_3\}$
- So we conclude that the DTMC does **not** satisfy the claimed property $P_{>0.99}[\text{try} \vee \neg\text{fail} \text{ } \mathbf{U}^{\leq 2} \text{ succ}]$.

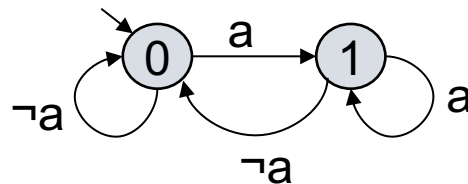
Buchi vs Rabin

- Consider the LTL property $\Diamond \Box a$. This can be described by this Buchi automaton, with $\{1\}$ as the accepting state:



Notice that this Buchi is **non-deterministic**. As such, we can use it for model checking on a probabilistic model such as DTMC.

- We can however represent the property with a **deterministic** Rabin automaton, with the pair $(\{0\}, \{1\})$ as its accepting condition.



additional slides MDP

Calculating $\text{pmin}(s, \varphi_1 \mathbf{U} \varphi_2)$

$\text{pmin}(s, \varphi)$ = the **minimum** probability for having executions starting from s satisfying φ , regardless the adversary.

To calculate $\text{pmin}(s, \varphi_1 \mathbf{U} \varphi_2)$:

1. Calculate first the sets S^{yes} and S^{no} .
 - $S^{\text{yes}} = \{ s \mid P(s, \varphi_1 \mathbf{U} \varphi_2) \geq 1, \text{ for all adversaries } \} \rightarrow \text{with algorithm Prob1A.}$
 - $S^{\text{no}} = \{ s \mid P(s, \varphi_1 \mathbf{U} \varphi_2) \leq 0, \text{ for some adversary } \} \rightarrow \text{with algorithm Prob0E.}$
2. For any state s in S^{yes} , we then know that $\text{pmin}(s, \varphi_1 \mathbf{U} \varphi_2) \geq 1$.
3. For any state s in S^{no} we have $\text{pmin}(s, \varphi_1 \mathbf{U} \varphi_2) \leq 0$.
4. We then proceed with calculating the pmin for the remaining states (which are not in S^{yes} nor S^{no}).

Algorithm Prob0E

- The algorithm below first calculates the set R of all states s satisfying $E[s, \varphi_1 \cup \varphi_2]$, regardless the adversary. So, for any state in R , and for any adversary $\text{Prob}(s, \varphi_1 \cup \varphi_2) > 0$.
- S^{no} is just complement S/R .
- $\text{Sat}(\varphi_1)$ and $\text{Sat}(\varphi_2)$ in the paremeters are the set of states on which φ_1 and φ_2 respectively hold.

$\text{PROB0E}(\text{Sat}(\phi_1), \text{Sat}(\phi_2))$

```
1.   $R := \text{Sat}(\phi_2)$ 
2.   $done := \text{false}$ 
3.  while ( $done = \text{false}$ )
4.       $R' := R \cup \{s \in \text{Sat}(\phi_1) \mid \forall \mu \in \text{Steps}(s). \exists s' \in R. \mu(s') > 0\}$ 
5.      if ( $R' = R$ ) then  $done := \text{true}$ 
6.       $R := R'$ 
7.  endwhile
8.  return  $S \setminus R$ 
```

Prob1A

- Calculate first the set F of states from where we have a path passing exclusively through states satisfying $\varphi_1 \wedge \neg\varphi_2$ and ends in S^{no} , under **some** adversary.
By definition this F also includes S^{no} .
- So any state s in F has $\text{Prob}(s, \varphi_1 \mathbf{U} \varphi_2) < 1$, for some adversary. In other words $\text{pmin}(s, \varphi_1 \mathbf{U} \varphi_2) < 1$.
- P^{yes} in the complement S/F .
- **Note:** for the calculation of $\text{pmin}(s, \varphi_1 \mathbf{U} \varphi_2)$, we can also just take $S^{\text{yes}} = \text{Sat}(\varphi_2)$. The calculation would still work, though it would take more steps to get its final results.

Calculating $\text{pmax}(s, \varphi_1 \mathbf{U} \varphi_2)$

$\text{pmax}(s, \varphi)$ = the **maximum** probability for having executions starting from s satisfying φ , regardless the adversary.

To calculate $\text{pmax}(s, \varphi_1 \mathbf{U} \varphi_2)$:

1. Calculate first the sets S^{yes} and S^{no} .
 - $S^{\text{yes}} = \{ s \mid P(s, \varphi_1 \mathbf{U} \varphi_2) \geq 1, \text{ for } \mathbf{some} \text{ adversaries} \} \rightarrow \text{with algorithm prob1E.}$
 - $S^{\text{no}} = \{ s \mid P(s, \varphi_1 \mathbf{U} \varphi_2) \leq 0, \text{ for } \mathbf{all} \text{ adversary} \} \rightarrow \text{with algorithm prob0A.}$
2. For any state s in S^{yes} , we then know that $\text{pmax}(s, \varphi_1 \mathbf{U} \varphi_2) \geq 1$.
3. For any state s in S^{no} we have $\text{pmax}(s, \varphi_1 \mathbf{U} \varphi_2) \leq 0$.
4. We then proceed with calculating the pmin for the remaining states (which are not in S^{yes} nor S^{no}).

Algorithm Prob0A

- The algorithm below first calculates the set R of all states s satisfying $E[s, \varphi_1 \cup \varphi_2]$, for some adversary. So, for any state in R , there is an adversary such that $\text{Prob}(s, \varphi_1 \cup \varphi_2) > 0$.
- S^{no} is just complement S/R .
- $\text{Sat}(\varphi_1)$ and $\text{Sat}(\varphi_2)$ in the parameters are the set of states on which φ_1 and φ_2 respectively hold.

$\text{PROB0A}(\text{Sat}(\phi_1), \text{Sat}(\phi_2))$

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1.   $R := \text{Sat}(\phi_2)$ 
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5.      if ( $R' = R$ ) then  $done := \text{true}$ 
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7.  endwhile
8.  return  $S \setminus R$ 
```

Prob1E

- More complicated. See Dave's slides on MDP.

Prob1E