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Introduction
Why do we need probabilistic systems?

- Our world is inherently stochastic:
  - Component failures
  - Inputs arrive with a probability distribution
  - Randomized algorithms / protocols

- Properties cannot be guaranteed to hold under all circumstances, but only with a certain probability.
  - The probability that component failures cause a safety-critical system state within 1000 steps is at most $10^{-6}$. 
Very simple model:

Discrete-time Markov Chains

\[ s_0 \xrightarrow{1.0} s_1 \xrightarrow{0.1} s_3 \]

\[ s_0 \xrightarrow{0.9} s_2 \xrightarrow{0.05} s_3 \]

\[ s_0 \xrightarrow{0.95} s_2 \xrightarrow{0.05} s_3 \]

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Extension to Computation Tree Logic (CTL)

- Path formulae as in CTL:
  - $X \varphi$
  - $\varphi_1 U \varphi_2$
  - $\varphi_1 U^{\leq k} \varphi_2$

- No $A$ and $E$ path quantifiers

- Instead: Probabilistic $\mathcal{P}_{\triangleright p}$ quantifier ($\triangleright \in \{\leq, \geq, <, >\}$)

$\mathcal{P}_{\triangleright p}(\psi)$: the probability to walk along a path satisfying $\psi$ has to be within the bound $\triangleright p$
Model Checking for PCTL: Traverse the syntax tree of the formula and mark the states with the subformulae that are satisfied there.

- $X$ 1 matrix-vector multiplication
- $U_{\leq k}$ $k$ matrix-vector multiplications
- $U$ Solution of a system of linear equations
  Reduction to matrix-vector multiplication by using iterative methods (e.g. Jacobi method)
A Discrete-time Markov Chain (DTMC)

Probability to compute:

\[ P^? \left( c \cup P_{\leq \frac{1}{2}}(a \cup b) \right) \]
Let $\gamma$ be a small constant $< \frac{1}{2}$.

<table>
<thead>
<tr>
<th>State</th>
<th>$P^?(a \cup b)$</th>
<th>$P_{\leq \frac{1}{2}}(a \cup b)$ satisfied?</th>
</tr>
</thead>
<tbody>
<tr>
<td>$s_1$</td>
<td>$\frac{1}{2} + \gamma^3$</td>
<td>no</td>
</tr>
<tr>
<td>$s_2$</td>
<td>$\frac{1}{2} + \gamma^3$</td>
<td>no</td>
</tr>
<tr>
<td>$s_3$</td>
<td>1</td>
<td>no</td>
</tr>
<tr>
<td>$s_4$</td>
<td>0</td>
<td>yes</td>
</tr>
<tr>
<td>$s_5$</td>
<td>$\gamma^2$</td>
<td>yes</td>
</tr>
<tr>
<td>$s_6$</td>
<td>$\gamma$</td>
<td>yes</td>
</tr>
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Let's see, what PRISM says for $\gamma = 10^{-6}$:

$$ \begin{align*}
\text{probabilistic } \\
\text{const double gamma = 0.000001;} \\
\text{module sys } \\
\text{ s: [1..6] init 1;} \\
\text{[] s=1 -> 1.0: (s'=2);} \\
\text{[] s=2 -> 0.5: (s'=3) + gamma: (s'=5) + (0.5-gamma): (s'=4);} \\
\text{[] s=3 -> 1.0: (s'=3);} \\
\text{[] s=4 -> 1.0: (s'=4);} \\
\text{[] s=5 -> gamma: (s'=6) + (1-gamma): (s'=4);} \\
\text{[] s=6 -> gamma: (s'=3) + (1-gamma): (s'=4);} \\
\text{endmodule}
\end{align*}$$

Result:

$$ \begin{align*}
\text{yes = 5, no = 1, maybe = 0} \\
\text{Time for model checking: 0.022 seconds.} \\
\text{Result: 1.0}
\end{align*} $$
What does MRMC say (for $\gamma = 10^{-6}$)?

Transitions:

```
STATES 6
TRANSITIONS 10
1 2 1.0
2 3 0.5
2 4 0.499999
2 5 0.000001
3 3 1.0
4 4 1.0
5 4 0.999999
5 6 0.000001
6 3 0.000001
6 4 0.999999
```

Labels:

```
#DECLARATION
a b c
#END
1 a
2 a
3 b
5 a
6 a
```

Result:

```
$RESULT: ( 1.0000000, 1.0000000, 0.0000000, 1.0000000, 1.0000000, 1.0000000 )
$STATE: { 1, 2, 4, 5, 6 }
```
The model checker has to represent the value $\frac{1}{2} + \gamma^3$ such that it is larger than 0.5.

For $\gamma = 10^{-6}$ this value is

$$\frac{1}{2} + 10^{-18} = 0.50000000000000001$$

Floating-point arithmetic with 64 bits can represent numbers with an accuracy of about 15 decimal digits.

The number is rounded downwards to 0.5.

Changing the value of $\frac{1}{2} + \gamma^3$ to 0.5 flips the truth value of $P_{\leq 0.5}(a \cup b)$ in $s_1$ and $s_2$ from “no” to “yes”.
Let $\gamma$ be a small constant $< \frac{1}{2}$.

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Sources of Inaccuracy
Sources of Inaccuracy

- **Inexact arithmetic** for additions and multiplications (floating-point arithmetic according to IEEE 754 standard)
- Termination criterion for the **iterative solution of linear equation systems** (Jacobi / Gauss-Seidel method)
- Symbolic Model Checking: **MTBDD-packages** only create a new leaf with value $v$ if there is no other leaf with value $v'$ and $|v - v'| < \varepsilon$ for some constant $\varepsilon > 0$.
- Continuous-time Markov chains: model checking for the time-bounded until operator requires **uniformization of the CTMC**, i.e. the evaluation of an infinite sum, which has to be truncated after a finite number of summands.
Ideas for Obtaining Reliable Results
Degree of Belief (1)

Given the PCTL formula

$$\varphi = P^{p}(\psi)$$

we define the **degree of believe**

$$db(\varphi) = \min_{s \in S} |Pr(s, \psi) - p|$$

If $db(\varphi)$ is close to 0 for some sub-formula $\varphi$ of our formula under consideration, the risk is high that wrong results are produced due to inexact computations.
Degree of Belief (2)

**Advantage:**
- Easy to integrate into existing tools.

**Disadvantage:**
- Does not provide any guarantees for the correctness.
Exact Arithmetic (1)

Pro solves the problem of rounding when using floating-point arithmetic.

Pro solves the problem of the restriction in the BDD packages

Contra inexact solutions to linear equations systems remain due to the iterative solution methods.

Contra the problem of uniformization of CTMCs also remains.

Contra the evaluation of the time-bounded next operator on CTMCs requires the computation of $e^{-\lambda t}$ for a $\lambda > 0$ which cannot be done exactly using rational arithmetic.
Disadvantages:

- very expensive w. r. t. runtime and memory consumption
- does not solve most of the problems.
Interval Arithmetic (1)

- Compute **safe intervals** which contain the probability with which a path formula holds in a state, i.e.

\[ l_\psi(s) = [a, b] \text{ such that } a \leq \Pr(s, \psi) \leq b \]

- **Three-valued interpretation** of PCTL:

\[
\begin{align*}
s \models \mathcal{P}_\bowtie p(\psi) & \iff \forall x \in l_\psi(s) : x \bowtie p \\
\neg s \models \mathcal{P}_\bowtie p(\psi) & \iff \exists x \in l_\psi(s) : x \nbowtie p
\end{align*}
\]

result unknown otherwise.

There are papers about obtaining such intervals for the solutions of linear equation systems and for the uniformization of CTMCs (Fox-Glynn approximation).
Interval Arithmetic (2)

**Advantage:**
- Intervals can be derived efficiently for all necessary operations.
- Results are really correct.

**Disadvantage:**
- Overhead of a (probably) small factor
- We have to live with “unknowns”.
A certificate is an output besides the yes/no answer which makes it easy to check that the decision of the tool is correct.

**Example: Linear Programming**

**Question:** Has $Ax = b, \ x \geq 0$ a satisfying solution?

**Yes:** Give a satisfying assignment.

**No:** Give a satisfying assignment for $y^T A \geq 0, \ y^T b < 0$, since exactly one of the two systems is satisfiable.
Certificates for DTMCs (1)

Property:

\[ \mathcal{P}_{>p}(\varphi_1 U \varphi_2) \]

Certificate:

A finite set of finite paths with probability mass \( > p \).

Example

Property \( \mathcal{P}_{>0.5}(aUb) \):
Advantages:

- easily verifiable using exact arithmetic.
- can guide the designer to find bugs in the design

Disadvantages:

- only applicable for formulae of the form $P_{>p}(aUb)$ without nested sub-formulae
- certificates may consist of a large number of paths.

Paper:
Han, Katoen – Counterexamples in Probabilistic Model Checking, TACAS 2007
Conclusion

- Reliable results are critical in probabilistic model checking.
- State-of-the-art model checkers use inexact computations. This may lead to wrong results.
- Interval arithmetic seems promising, but further research is necessary.