Interactive verification of Markov chains: Two distributed protocol case studies

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QFM 2012 28 August 2012



In the interactive theorem prover



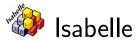
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 - ZeroConf protocol (IPv4 address allocation)



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- Built on Isabelle's probability theory and Markov chains Hölzl & Heller (ITP 2011), Hölzl & Nipkow (TACAS 2012)

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- Proof language and proof methods



► Logic is HOL: functional programming + quantifiers

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- Small kernel: each proof is reduced to primitive proof steps
- Powerful proof methods (rewrite engine, Sledgehammer, ...)
- Important theories: datatypes, real analysis, measure theory, probability theory, Markov chains, ...

Case study: ZeroConf protocol

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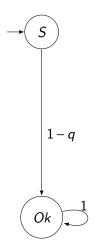
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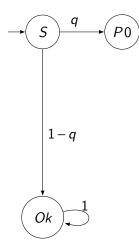
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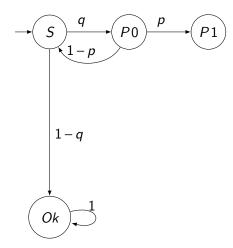
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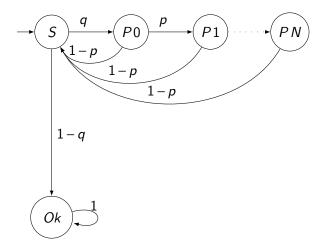
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- Model checking analysis of Kwiatkowska et al. (2006) and Andova et al. (2003)

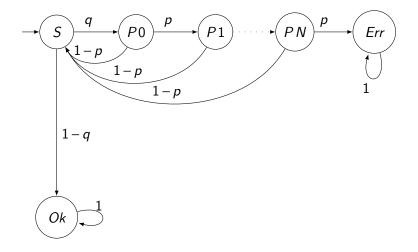


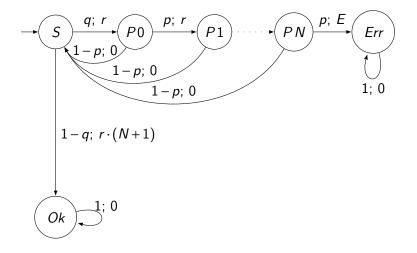


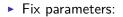




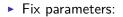




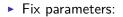




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Define state space:

datatype zc- $state = S | P \mathbb{N} | Ok | Err$

$$\Omega = \left\{ S, Ok, Err \right\} \cup \left\{ P n \mid n \le N \right\}$$

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Define the transition function τ:

$$\tau S \quad Ok = 1-q \tau S \quad (P \ 0) = q \tau (P \ n) (P(n+1)) = \text{ if } n < N \text{ then } p \text{ else } 0$$

• Defines a Markov chain:

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Analyse: Perr S =?

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case 0
show
$$P_{err} (P(N-0)) = p^{0+1} + (1-p^{0+1}) \cdot P_{err} S$$

by simp

$$\begin{split} n &\leq N \implies \mathsf{P}_{err} \ (P(N-n)) = p^{n+1} + (1-p^{n+1}) \cdot \mathsf{P}_{err} \ S \\ \text{proof } (induct \ n) \\ & \text{case } (n+1) \\ & \text{have } \mathsf{P}_{err} \ (P(N-(n+1))) \\ & = p \cdot (p^{n+1} + (1-p^{n+1}) \cdot \mathsf{P}_{err} \ S) + (1-p) \cdot \mathsf{P}_{err} \ S \\ & \text{by } (simp \cdots) \\ & \text{also have } \dots = p^{(n+1)+1} + (1-p^{(n+1)+1}) \cdot \mathsf{P}_{err} \ S \\ & \text{by } (simp \cdots) \\ & \text{finally show } \mathsf{P}_{err} \ (P(N-(n+1))) \\ & = p^{(n+1)+1} + (1-p^{(n+1)+1}) \cdot \mathsf{P}_{err} \ S \ . \\ & \text{next} \\ & \text{case } 0 \\ & \text{show } \mathsf{P}_{err} \ (P(N-0)) = p^{0+1} + (1-p^{0+1}) \cdot \mathsf{P}_{err} \ S \\ & \text{by } simp \end{split}$$

qed

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▶ 16 hosts (q = 16/65024), 3 probe runs (N = 2), p = 0.01:

corollary $P_{err} S \le 10^{-13}$

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- Similar to τ define the cost function ρ :

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▶ 16 hosts, 3 probe runs, p = 0.01, r = 2ms, E = 3600s:

theorem
$$C_{fin} S \le 0.007$$

Case study: Crowds protocol



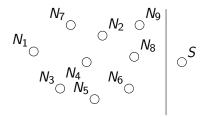
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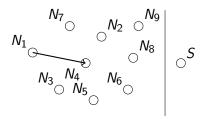
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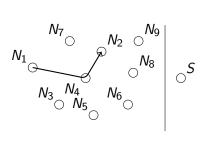
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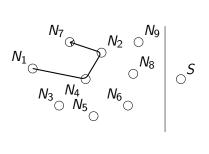
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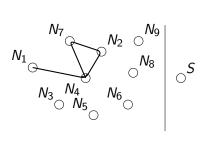
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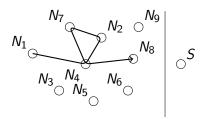


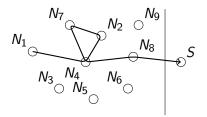


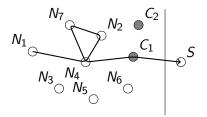


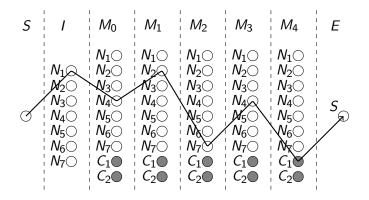








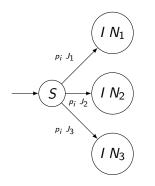


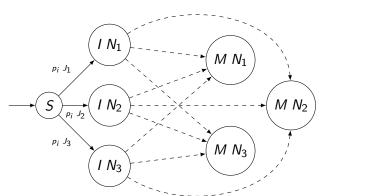


Probabilities:

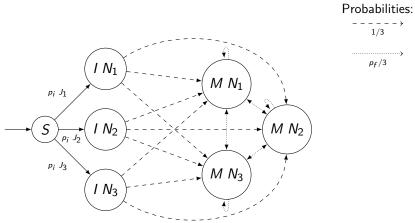


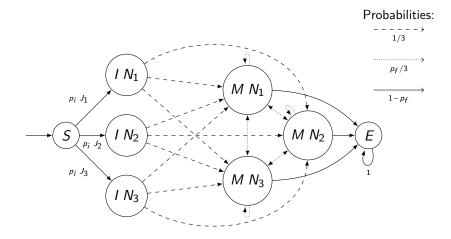
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Define state space

datatype α *c-state* = $S \mid I \mid \alpha \mid M \mid \alpha \mid E$

$$\Omega = \left\{ S \right\} \cup \left\{ I \ n \ \middle| \ n \in N \setminus C \right\} \cup \left\{ M \ n \ \middle| \ n \in N \right\} \cup \left\{ E \right\}$$

Define transition function

$$\begin{array}{l} \tau \ S & (I \ n) &= p_i \ n \\ \tau \ (I \ n) & (M \ n') &= 1/|N| \\ \tau \ (M \ n) \ (M \ n') &= p_f/|N| \\ \tau \ (M \ n) \ E &= 1 - p_f \\ \tau \ E & E &= 1 \\ \tau \ _ &= 0 \end{array}$$

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Prove Markov chain property

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Information the collaborating nodes gain when contacted

theorem
$$I_{hit}(init; last-ncoll) \le \left(1 - \frac{|N \setminus C| - 1}{|N|} \cdot p_f\right) \cdot \log_2 |N \setminus C|$$

Related Work: probability theory in ITPs

Probability space of boolean sequences: N→ {0,1}
 Hurd (2002), Hasan *et al.* (2009), Liu *et al.* (2011)

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- Formalization of pGCL (prob. & non-det. language)
 Hurd *et al.* (2005), Audebaud & Paulin-Mohring (2009)

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Future Work:

- More Markov models (MDPs, CTMCs, CTMDPs, PTAs)
- Certification of probabilistic model checker results
- Specification language

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Slides available at: http://www.in.tum.de/~hoelzl