

Quantitative Verification: Correctness, Reliability and Beyond



Dave Parker
University of Birmingham

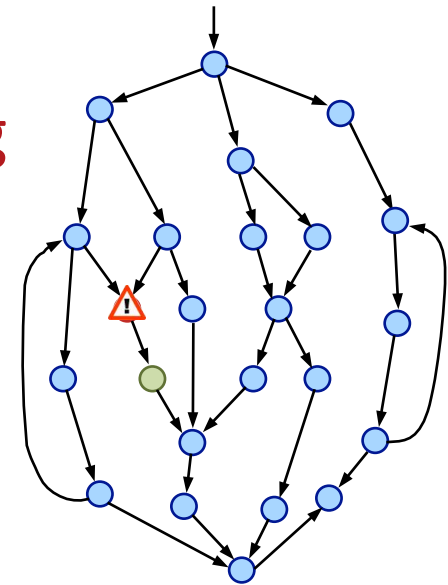
December 2012

Verification

- Checking the correctness of (computerised) systems using rigorous, mathematically-sound techniques
 - in essence: **proving** that a piece of software, or hardware, or a protocol behaves correctly

- Automated verification: **model checking**

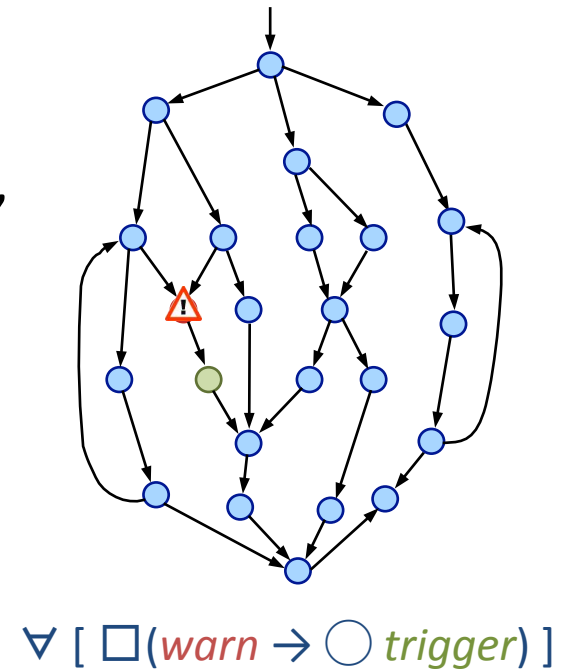
- correctness properties expressed in temporal logic
- exhaustive construction/analysis of finite-state model



$\forall [\square (warn \rightarrow \bigcirc trigger)]$

Model checking

- Successful in practice
 - e.g. Windows device drivers, circuit designs, ...
- Example properties
 - “acquire/release of spinlock is always done in strict alternation”
 - “no array is accessed outside its bounds”
- Why it works
 - temporal logic: expressive, tractable
 - fully automated, tools available
 - not just verification, but falsification of properties, i.e. bug hunting



Quantitative verification

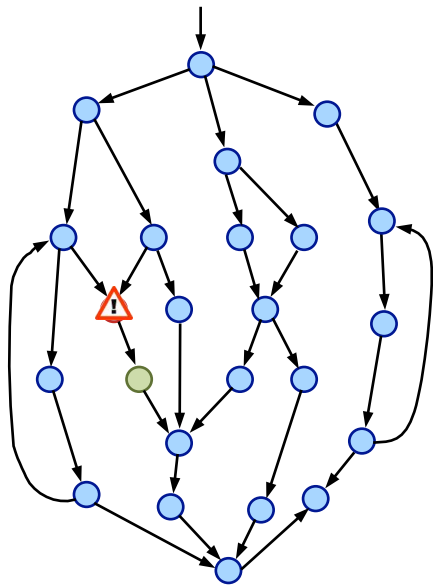
- Adds quantitative aspects (to models and properties)
 - probability, time, costs, rewards, ...
- Probability
 - physical components can fail
 - communication media are unreliable
 - algorithms/protocols use randomisation
- Time
 - delays, time-outs, failure rates, ...
- Costs & rewards
 - power consumption, resource usage, ...
 - profit, incentive schemes, ...



Probabilistic model checking

- Construction and analysis of probabilistic models
 - Markov chains, Markov decision processes, ...
- Correctness properties in probabilistic temporal logic
 - $P_{>0.999} [\Box(\text{trigger} \rightarrow \Diamond^{\leq 20} \text{deploy})]$
 - “the probability of an airbag always deploying within 20ms of being triggered is at least 0.999”
 - correctness, reliability, performance, ...
- Model checking algorithms (and tools)
 - graph algorithms, linear equations, linear programming, numerical fixed points, numerical approximations, ...

Probabilistic models



+ probabilities



Markov
chain

e.g. communic-
ation protocol

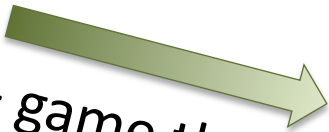
+ exponential
time delays



continuous-
time Markov
chain

e.g. systems
biology

+ game theory

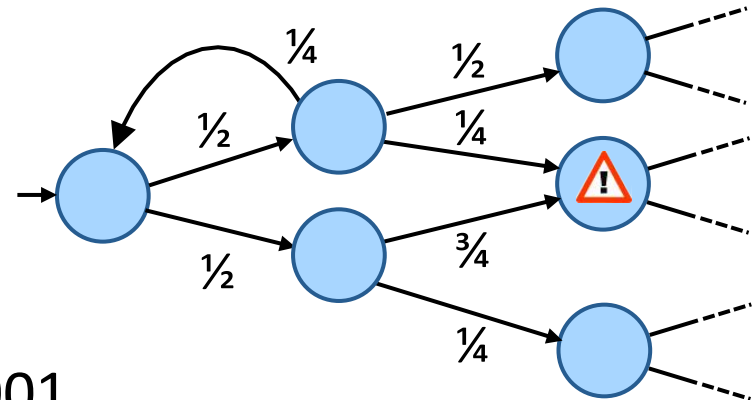


stochastic
game

e.g. energy
management

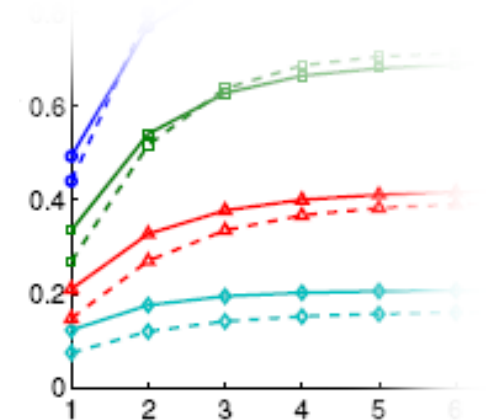
1) Adding: Probabilities

- Model: Markov chain
 - add probabilities to transitions



- Properties
 - probability of airbag failure < 0.001
 - **numerical** queries: what is the probability of failure?

- Key ideas:
 - **exact** numerical results
 - combines **numerical** + **exhaustive** analysis
 - results show system flaws, anomalies

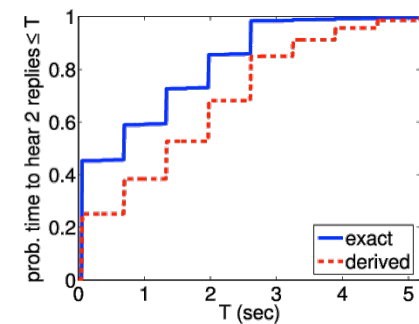


- Applications
 - network protocols, security, biology, robotics & planning, power management, nanotechnology...

Example: Bluetooth

- Device discovery between a pair of Bluetooth devices
 - performance essential for this phase
- Complex discovery process
 - two asynchronous 28-bit clocks
 - pseudo-random hopping between 32 frequencies
 - random waiting scheme to avoid collisions
 - 17,179,869,184 initial configurations
- Probabilistic model checking
 - “worst-case expected discovery time is at most 5.17s”
 - “probability discovery time exceeds 6s is always < 0.001 ”

$$\text{freq} = [\text{CLK}_{16-12} + k + (\text{CLK}_{4-2,0} - \text{CLK}_{16-12}) \bmod 16] \bmod 32$$



2) Adding: Exponential delays

- Continuous-time Markov chains

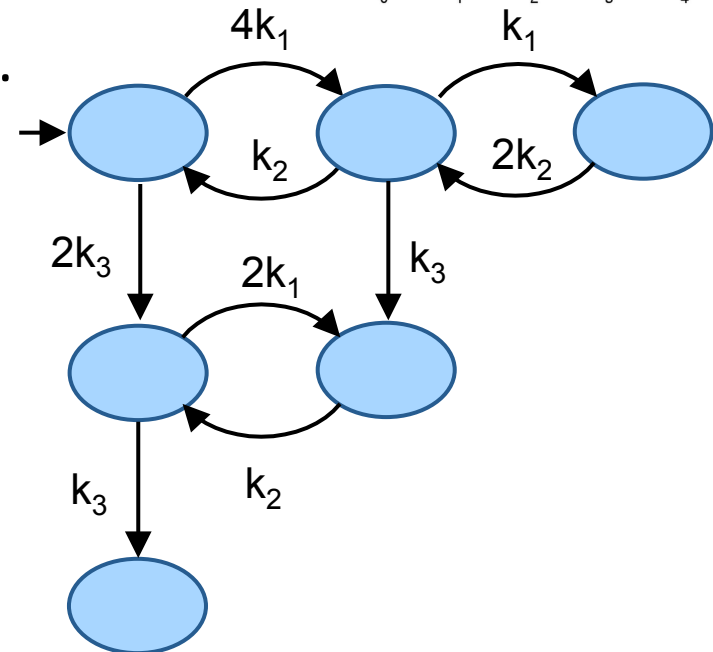
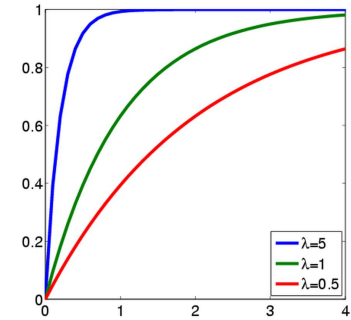
- random delays on transitions between states
- delays are exponentially distributed
- e.g. failure rates, reaction times, ...

- Applications

- network performance models
- biological reactions

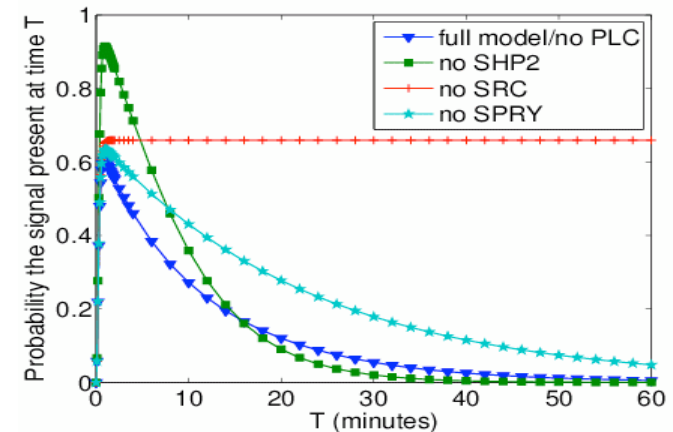
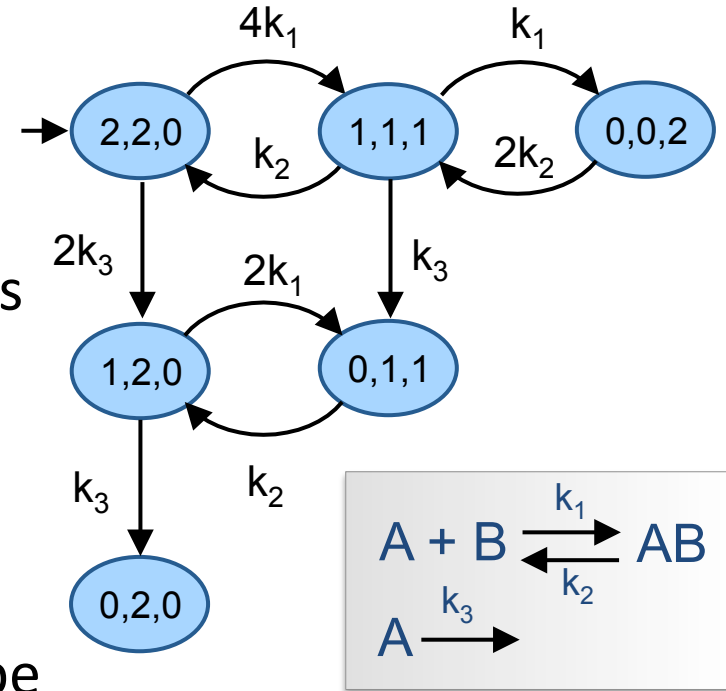
- Properties

- probability of disk-failure within 1 month?
- expected number of molecules of X at time instant T?



Example: Systems biology

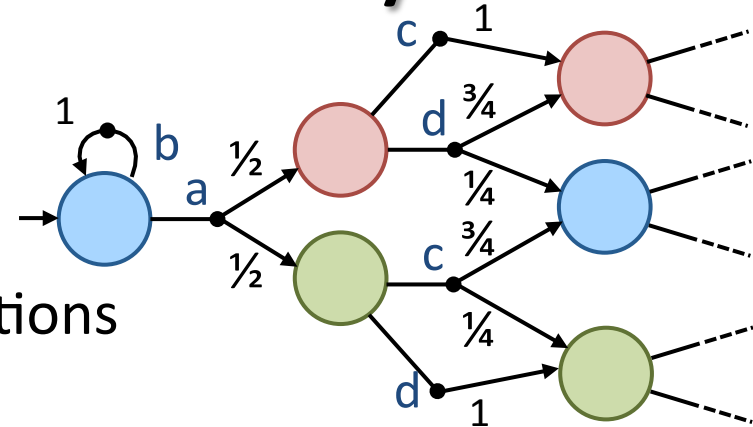
- Markov model of reactions
 - states represent molecule counts
 - transitions correspond to reactions
- Key ideas
 - “in-silico” experiments
 - aim: validate biologists’ models
 - probabilistic model checking can be cheaper than simulation
 - small models yield useful results
- Case study: FGF pathway
 - model developed with biologists
 - validated against lab experiments



3) Adding: Game theory

- Multi-player stochastic games

- states **controlled** by **players**
- players choose (probabilistic) actions



- Key ideas

- automated methods essential to reason about complex player strategies, and interaction with probabilities

- Property specifications

- does player 1 have a strategy to ensure that the probability of is < 0.01 , regardless of the strategies if players 2 and 3?

- Applications

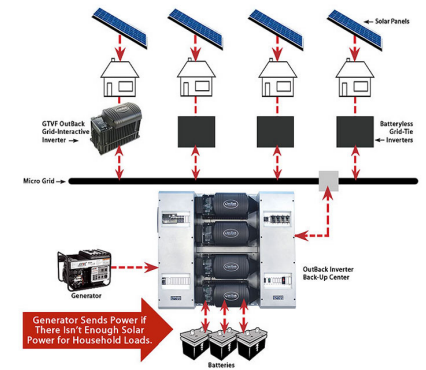
- controller synthesis (controller vs. environment), security (system vs. attacker), distributed algorithms, ...

Example: Energy management

- Energy management protocol for Microgrid

- Microgrid: local energy management
- randomised demand management protocol
- probability: randomisation, demand model, ...

[Hildmann/
Saffre'11]

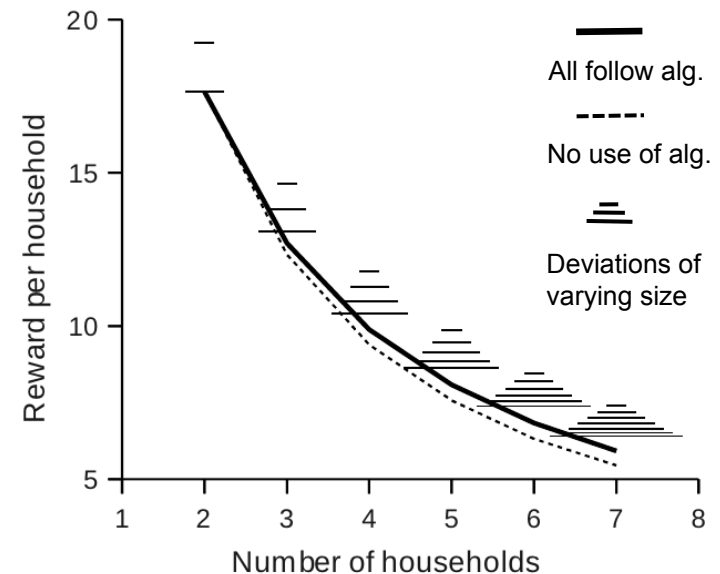


- Existing analysis

- simulation-based
- assumes all clients are unselfish

- Our analysis

- stochastic multi-player game
- clients can cheat (and cooperate)
- exposes protocol weakness
- propose/verify simple fix



Conclusions

- Quantitative verification
 - formal methods to build/analyse probabilistic models
 - temporal logics for correctness, reliability, performance, ...
 - exact results, combines numerical + exhaustive analysis
 - wide range of applications
- Challenges
 - scalability + efficiency
 - wider property classes, e.g. partial information for games
 - richer models: timed games, hybrid automata, ...