



Automatic Verification of Competitive Stochastic Systems

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Verifying stochastic systems

- **Quantitative verification**
 - probability, time, costs/rewards, ...
 - in particular: systems with stochastic behaviour
 - e.g. due to unreliability, uncertainty, randomisation, ...
 - often: subtle interplay between probability/nondeterminism
- **Automated verification**
 - probabilistic model checking
 - tool support: PRISM model checker
 - techniques for improving efficiency, scalability
- **Practical applications**
 - wireless communication protocols, security protocols, systems biology, DNA computing, robotic planning, ...

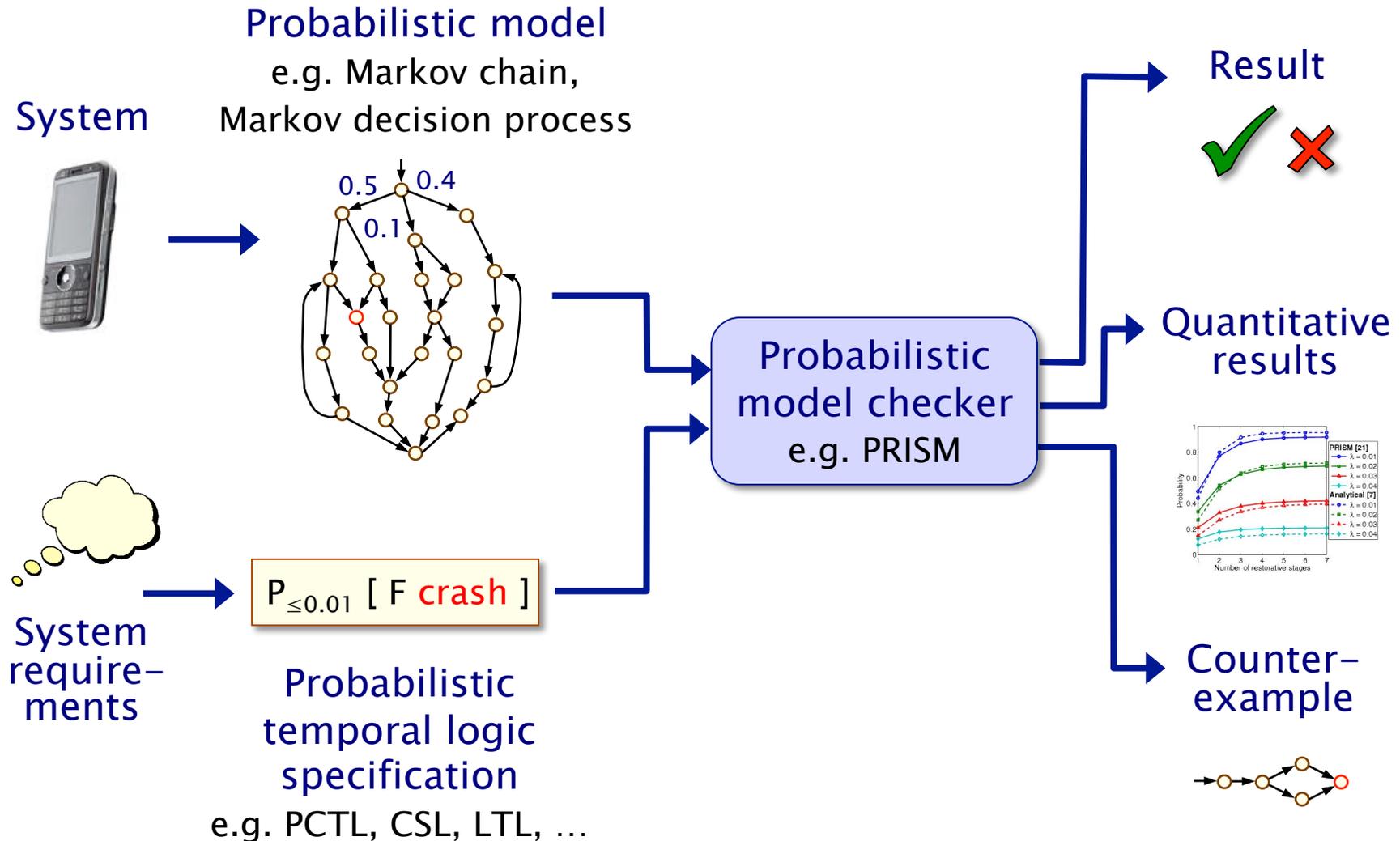
Adding competitive behaviour

- **Open systems**
 - need to account for the behaviour of components not under our control, possibly with differing/opposing goals
 - giving rise to competitive behaviour
- **Many occurrences in practice**
 - e.g. security protocols, algorithms for distributed consensus, energy management or sensor network co-ordination
- **Natural to adopt a game-theoretic view**
 - widely used in computer science, economics, ...
- **This talk**
 - verifying systems with **competitive** and **stochastic** behaviour
 - stochastic multi-player games
 - temporal logic, model checking, tool support, case studies

Overview

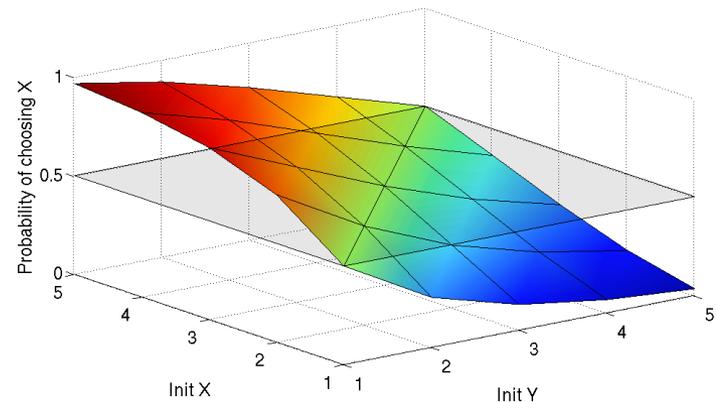
- Probabilistic model checking
 - probabilistic models, property specifications
- Stochastic multi-player games (SMGs)
 - the model, probability spaces, rewards
- Property specification: rPATL
 - syntax, semantics, subtleties
- rPATL model checking
 - algorithm, numerical computation, tool support
- Case study: energy management in microgrids

Probabilistic model checking



Probabilistic model checking

- Property specifications based on temporal logic
 - PCTL, CSL, probabilistic LTL, PCTL*, ...
- Simple examples:
 - $P_{\leq 0.01} [F \text{ “crash” }]$ – “the probability of a crash is at most 0.01”
 - $S_{>0.999} [\text{“up”}]$ – “long-run probability of availability is >0.999 ”
- Usually focus on **quantitative** (numerical) properties:
 - $P_{=?} [F \text{ “crash” }]$
“what is the probability of a crash occurring?”
 - then analyse trends in quantitative properties as system parameters vary

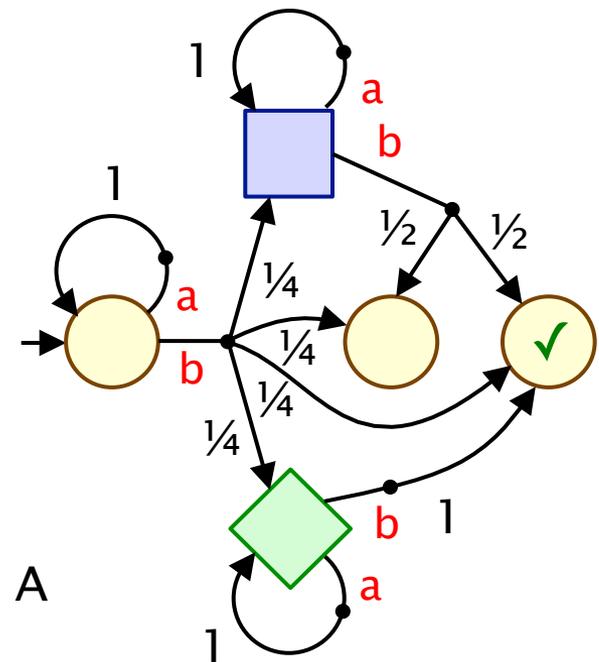


Probabilistic model checking

- Typically combine **numerical** + **exhaustive** aspects
 - $P_{\max=?} [F^{\leq 10} \text{“fail”}]$ – “worst-case probability of a failure occurring within 10 seconds, for any possible scheduling of system components”
 - $P_{=?} [G^{\leq 0.02} !\text{“deploy”} \{ \text{“crash”} \}^{\max}]$ – “the maximum probability of an airbag failing to deploy within 0.02s, from any possible crash scenario”
 - model checking: graph analysis + numerical solution + ...
- **Reward-based properties (rewards = costs = prices)**
 - $R_{\{\text{“time”}\}=?} [F \text{“end”}]$ – “expected algorithm execution time”
 - $R_{\{\text{“energy”}\}_{\max=?}} [C^{\leq 7200}]$ – “worst-case expected energy consumption during the first 2 hours”

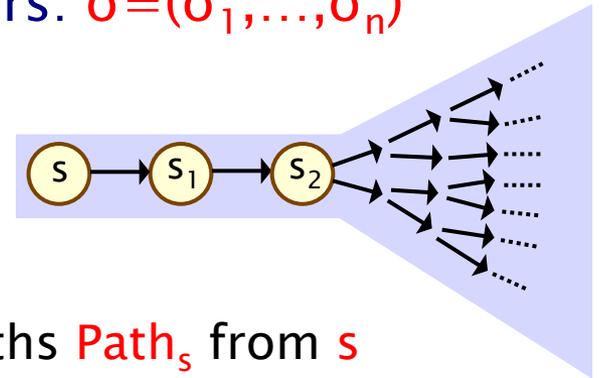
Stochastic multi-player games

- Stochastic multi-player game (SMGs)
 - probability + nondeterminism + multiple players
- A (turn-based) SMG is a tuple $(\Pi, S, \langle S_i \rangle_{i \in \Pi}, A, \Delta, L)$:
 - Π is a set of n players
 - S is a (finite) set of states
 - $\langle S_i \rangle_{i \in \Pi}$ is a partition of S
 - A is a set of action labels
 - $\Delta : S \times A \rightarrow \text{Dist}(S)$ is a (partial) transition probability function
 - $L : S \rightarrow 2^{\text{AP}}$ is a labelling with atomic propositions from AP
- Notation:
 - $A(s)$ denotes available actions in state s



Paths, strategies + probabilities

- **Path**: is an (infinite) sequence of connected states in SMG
 - represents a system execution (i.e. one possible behaviour)
- **Strategy** for player i : resolves choices in S_i states
 - based on execution history, i.e. $\sigma_i : (SA)^*S_i \rightarrow \text{Dist}(A)$
 - Σ_i denotes the set of all strategies for player i
- **Strategy profile**: strategies for all players: $\sigma = (\sigma_1, \dots, \sigma_n)$
 - can be: deterministic (pure), memoryless, finite-memory, ...
- **Probability measure over paths**: \Pr_s^σ
 - for strategy profile σ , over set of all paths Path_s from s
 - any (ω -)regular property over states/actions is measurable
 - $E_s^\sigma [X]$: expected value of measurable function $X : \text{Path}_s \rightarrow \mathbb{R}_{\geq 0}$



Rewards

- **Rewards** (or costs)
 - real-valued quantities assigned to states (and/or transitions)
- **Wide range of possible uses:**
 - elapsed time, energy consumption, size of message queue, number of messages successfully delivered, net profit, ...
- **We use:**
 - state rewards: $r : S \rightarrow \mathbb{N}$ (but can generalise to $\mathbb{Q}_{\geq 0}$)
 - **expected cumulative** reward until a target set **T** is reached
- **3 interpretations of rewards**
 - 3 reward types $* \in \{\infty, c, 0\}$, differing where T is not reached
 - reward is assumed to be infinite, cumulated sum, zero, resp.
 - ∞ : e.g. expected time for algorithm execution
 - **c**: e.g. expected resource usage (energy, messages sent, ...)
 - **0**: e.g. reward incentive awarded on algorithm completion

Property specification: rPATL

- New temporal logic **rPATL**:
 - reward probabilistic alternating temporal logic
- CTL, extended with:
 - coalition operator $\langle\langle C \rangle\rangle$ of ATL
 - probabilistic operator **P** of PCTL
 - generalised version of reward operator **R** from PRISM
- Example:
 - $\langle\langle \{1,2\} \rangle\rangle P_{<0.01} [F^{\leq 10} \text{error}]$
 - “players 1 and 2 have a strategy to ensure that the probability of an error occurring within 10 steps is less than 0.01, regardless of the strategies of other players”

rPATL syntax

- Syntax:

$$\begin{aligned}\phi &::= \top \mid a \mid \neg\phi \mid \phi \wedge \phi \mid \langle\langle C \rangle\rangle P_{\bowtie q}[\psi] \mid \langle\langle C \rangle\rangle R_{\bowtie x}^r [F^* \phi] \\ \psi &::= X \phi \mid \phi U^{\leq k} \phi \mid F^{\leq k} \phi \mid G^{\leq k} \phi\end{aligned}$$

- where:

- $a \in AP$ is an atomic proposition, $C \subseteq \Pi$ is a coalition of players,
 $\bowtie \in \{\leq, <, >, \geq\}$, $q \in [0, 1] \cap \mathbb{Q}$, $x \in \mathbb{Q}_{\geq 0}$, $k \in \mathbb{N} \cup \{\infty\}$
 r is a reward structure and $^* \in \{0, \infty, c\}$ is a reward type

- $\langle\langle C \rangle\rangle P_{\bowtie q}[\psi]$

- “players in coalition C have a strategy to ensure that the probability of path formula ψ being true satisfies $\bowtie q$, regardless of the strategies of other players”

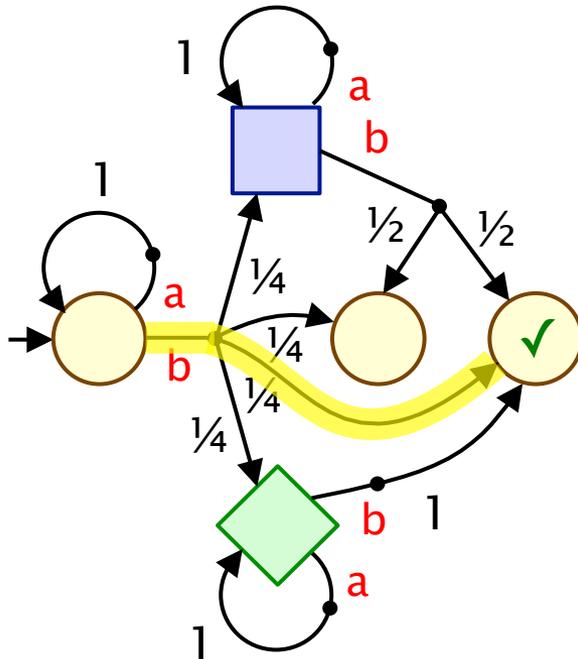
- $\langle\langle C \rangle\rangle R_{\bowtie x}^r [F^* \phi]$

- “players in coalition C have a strategy to ensure that the expected reward r to reach a ϕ -state (type *) satisfies $\bowtie x$, regardless of the strategies of other players”

rPATL semantics

- Semantics for most operators is standard
- Just focus on P and R operators...
 - present using reduction to a stochastic 2-player game
 - (as for later model checking algorithms)
- Coalition game G_C for SMG G and coalition $C \subseteq \Pi$
 - 2-player SMG where C and $\Pi \setminus C$ collapse to players 1 and 2
- $\langle\langle C \rangle\rangle P_{\bowtie q}[\psi]$ is true in state s of G iff:
 - in coalition game G_C :
 - $\exists \sigma_1 \in \Sigma_1$ such that $\forall \sigma_2 \in \Sigma_2 . \Pr_s^{\sigma_1, \sigma_2}(\psi) \bowtie q$
- Semantics for R operator defined similarly...

Examples



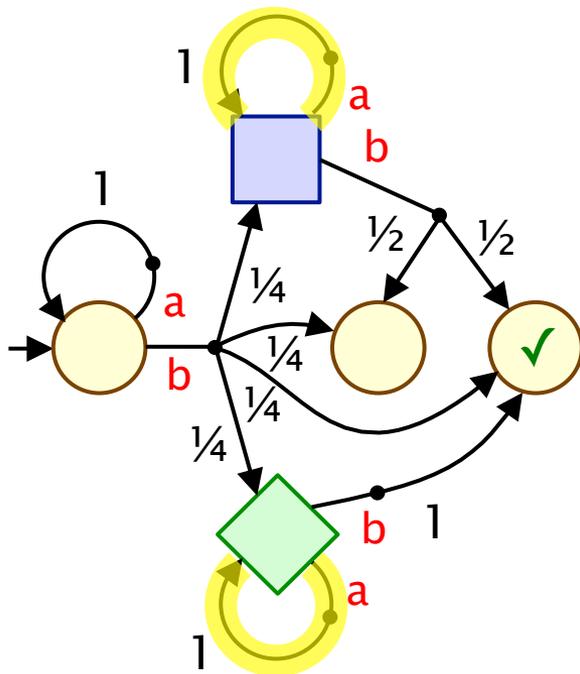
$$\langle\langle \bigcirc \rangle\rangle P_{\geq 1/4} [F \checkmark]$$

true in initial state

$$\langle\langle \bigcirc \rangle\rangle P_{\geq 1/3} [F \checkmark]$$

$$\langle\langle \bigcirc, \square \rangle\rangle P_{\geq 1/3} [F \checkmark]$$

Examples



$$\langle\langle \bigcirc \rangle\rangle P_{\geq 1/4} [F \checkmark]$$

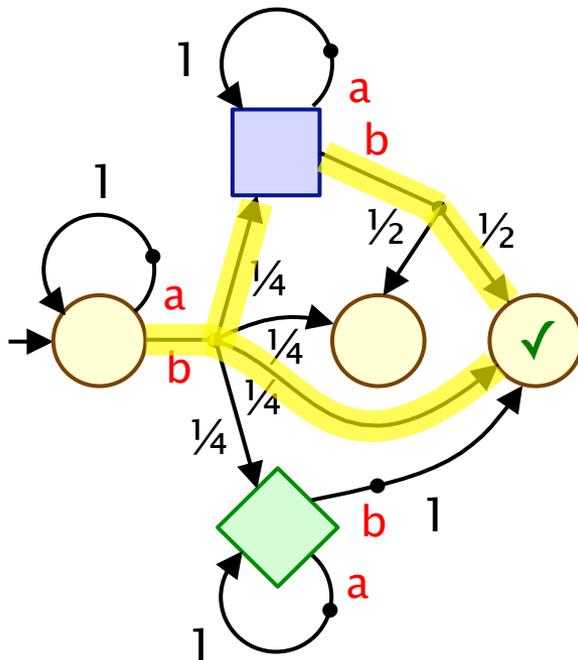
true in initial state

$$\langle\langle \bigcirc \rangle\rangle P_{\geq 1/3} [F \checkmark]$$

false in initial state

$$\langle\langle \bigcirc, \square \rangle\rangle P_{\geq 1/3} [F \checkmark]$$

Examples



$$\langle\langle \text{○} \rangle\rangle P_{\geq 1/4} [F \checkmark]$$

true in initial state

$$\langle\langle \text{○} \rangle\rangle P_{\geq 1/3} [F \checkmark]$$

false in initial state

$$\langle\langle \text{○}, \text{□} \rangle\rangle P_{\geq 1/3} [F \checkmark]$$

true in initial state

Equivalences + extensions

- Two useful equivalences:
- $\langle\langle C \rangle\rangle P_{\geq q}[\neg\psi] \equiv \langle\langle C \rangle\rangle P_{\leq 1-q}[\psi]$
 - **negation** to derive path properties e.g. $G a \equiv \neg F \neg a$
 - model checking essentially just focuses on reachability
- $\langle\langle C \rangle\rangle P_{\geq q}[\psi] \equiv \neg \langle\langle \Pi \setminus C \rangle\rangle P_{< q}[\psi]$
 - thanks to standard **determinacy** results
 - model checking focuses on min/max values for P1/P2
- Quantitative (**numerical**) properties:
 - best/worst-case values
- e.g. $\langle\langle C \rangle\rangle P_{\max=?}[\psi] = \sup_{\sigma_1 \in \Sigma_1} \inf_{\sigma_2 \in \Sigma_2} \Pr_s^{\sigma_1, \sigma_2}(\psi)$

Independence of strategies

- Strategies for each coalition operator are independent
 - for example, in: $\langle\langle 1 \rangle\rangle P_{\geq 1} [G (\langle\langle 1, 2 \rangle\rangle P_{\geq \frac{1}{4}} [F \checkmark])]$
 - no dependencies in player 1 strategies in quantifiers
 - branching-time temporal logic (like ATL, PCTL, ...)
- Introducing dependencies is problematic
 - e.g. subsumes existential semantics for PCTL on Markov decision processes (MDPs), which is undecidable
 - (does there exist a single adversary satisfying one formula?)
 - $\langle\langle 1 \rangle\rangle P_{\geq 1} [G \langle\langle 1 \rangle\rangle P_{\geq \frac{1}{4}} [F \checkmark]]$
- But nested properties still have natural applications
 - e.g. sensor network, with players: **sensor**, **repairer**
 - $\langle\langle \text{sensor} \rangle\rangle P_{< 0.01} [F (\neg \langle\langle \text{repairer} \rangle\rangle P_{\geq 0.95} [F \text{“operational”}])]$

Why do we need multiple players?

- SMGs have multiple (>2) players
 - but semantics (and model checking) reduce to 2-player case
 - due to (zero sum) nature of queries expressible by rPATL
 - so why do we need multiple players?
- 1. Modelling convenience
 - and/or multiple rPATL queries on same model
- 2. May also exploit in nested queries, e.g.:
 - players: sensor1, sensor2, repairer
 - $\langle\langle \text{sensor1} \rangle\rangle P_{<0.01} [F (\neg \langle\langle \text{repairer} \rangle\rangle P_{\geq 0.95} [F \text{ “operational” }])]$

Model checking rPATL

- Basic algorithm: as for any branching-time temporal logic
 - recursive descent of formula parse tree
 - compute $\text{Sat}(\phi) = \{ s \in S \mid s \models \phi \}$ for each subformula ϕ
- Main task: checking P and R operators
 - reduction to solution of stochastic 2-player game G_C
 - e.g. $\langle\langle C \rangle\rangle P_{\geq q}[\psi] \Leftrightarrow \sup_{\sigma_1 \in \Sigma_1} \inf_{\sigma_2 \in \Sigma_2} \Pr_s^{\sigma_1, \sigma_2}(\psi) \geq q$
 - complexity: $\text{NP} \cap \text{coNP}$ (without any $R[F^0]$ operators)
 - compared to, e.g. P for Markov decision processes
 - complexity for full logic: $\text{NEXP} \cap \text{coNEXP}$ (due to $R[F^0]$ op.)
- In practice though:
 - evaluation of numerical **fixed points** (“value iteration”)
 - up to a desired level of convergence
 - usual approach taken in probabilistic model checking tools

Probabilities for P operator

- E.g. $\langle\langle C \rangle\rangle P_{\geq q}[F \phi]$: max/min reachability probabilities
 - compute $\sup_{\sigma_1 \in \Sigma_1} \inf_{\sigma_2 \in \Sigma_2} \Pr_s^{\sigma_1, \sigma_2}(F \phi)$ for all states s
 - deterministic memoryless strategies suffice

- Value is:

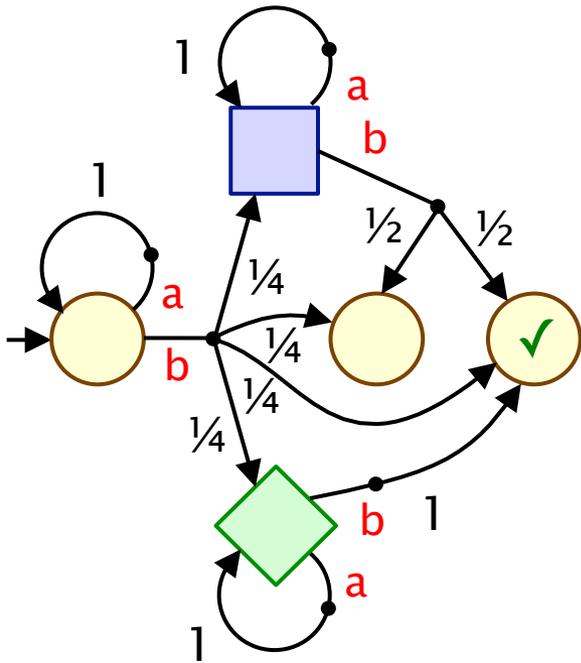
- 1 if $s \in \text{Sat}(\phi)$, and otherwise **least** fixed point of:

$$f(s) = \begin{cases} \max_{a \in A(s)} \left(\sum_{s' \in S} \Delta(s, a)(s') \cdot f(s') \right) & \text{if } s \in S_1 \\ \min_{a \in A(s)} \left(\sum_{s' \in S} \Delta(s, a)(s') \cdot f(s') \right) & \text{if } s \in S_2 \end{cases}$$

- Computation:

- start from zero, propagate probabilities backwards
- guaranteed to converge

Example



rPATL: $\langle\langle \text{yellow circle}, \text{blue square} \rangle\rangle P_{\geq 1/3} [F \checkmark]$

Player 1: yellow circle, blue square Player 2: green diamond

Compute: $\sup_{\sigma_1 \in \Sigma_1} \inf_{\sigma_2 \in \Sigma_2} \Pr_s^{\sigma_1, \sigma_2} (F \checkmark)$

Rewards for $R[F^c]$ operator

- E.g. $\langle\langle C \rangle\rangle R_{\geq q}^r [F^c \phi]$: max/min expected rewards for P1 /P2
 - again: deterministic memoryless strategies suffice
- Value is:
 - ∞ if $s \in \text{Sat}(\langle\langle C \rangle\rangle P_{>0} [G F \text{ "pos_rew" }])$,
 - 0 if $s \in \text{Sat}(\phi)$, and otherwise **least** fixed point of:

$$f(s) = \begin{cases} r(s) + \max_{a \in A(s)} \left(\sum_{s' \in S} \Delta(s, a)(s') \cdot f(s') \right) & \text{if } s \in S_1 \\ r(s) + \min_{a \in A(s)} \left(\sum_{s' \in S} \Delta(s, a)(s') \cdot f(s') \right) & \text{if } s \in S_2 \end{cases}$$

Rewards for $R[F^\infty]$ operator

- E.g. $\langle\langle C \rangle\rangle R_{\geq q}^r [F^\infty \phi]$: max/min expected rewards for P1 / P2
 - again: deterministic memoryless strategies suffice

- Value is:

- ∞ if $s \in \text{Sat}(\langle\langle C \rangle\rangle P_{>0} [G F \text{ "pos_rew" }])$,
- 0 if $s \in \text{Sat}(\phi)$, and otherwise **greatest** fixed point **over** \mathbb{R} of:

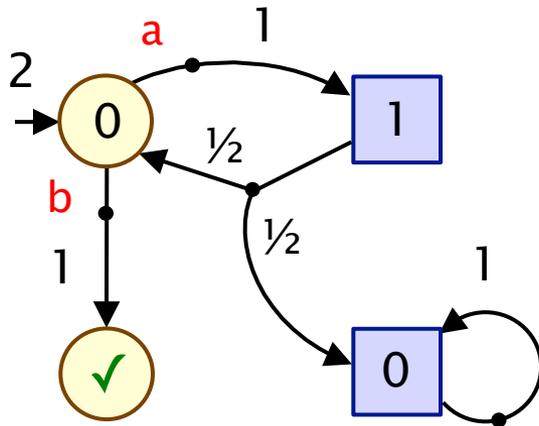
$$f(s) = \begin{cases} r(s) + \max_{a \in A(s)} \left(\sum_{s' \in S} \Delta(s, a)(s') \cdot f(s') \right) & \text{if } s \in S_1 \\ r(s) + \min_{a \in A(s)} \left(\sum_{s' \in S} \Delta(s, a)(s') \cdot f(s') \right) & \text{if } s \in S_2 \end{cases}$$

- Computation:

- 1. set zero rewards to ϵ , compute least fixed point
- 2. evaluate greatest fixed point, downwards from step 1

Example: Finite memory for R[F0]

- E.g. $\langle\langle C \rangle\rangle R_{\geq q}^r [F^0 \phi]$: max/min expected rewards for P1 / P2
 - now: deterministic memoryless strategies **do not** suffice



$$\langle\langle \text{○}, \text{□} \rangle\rangle R_{\geq \frac{1}{2}}^r [F^0 \checkmark]$$

- b**: reward 0
- a**, **b**: expected reward 0.5
- a**, **a**, **b**: expected reward 0.5
- a**, **a**, **a**, **b**: expected reward 0.375

What if incoming reward is 2?

- b**: reward 2
- a**, **b**: expected reward 1.5

Rewards for $R[F^0]$ operator

- E.g. $\langle\langle C \rangle\rangle R_{\geq q}^r [F^0 \phi]$: max/min expected rewards for P1 /P2
 - now: deterministic memoryless strategies **do not** suffice
- There exists a **finite-memory** optimal strategy for P1
 - there exists a bound B, beyond which strategy is memoryless
 - B is exponential in worst-case, but can be computed...
- **Computation:**
 - compute bound B (using simpler rPATL queries)
 - perform value iteration for each level $0, \dots, B$; combine results

Tool support: PRISM-games



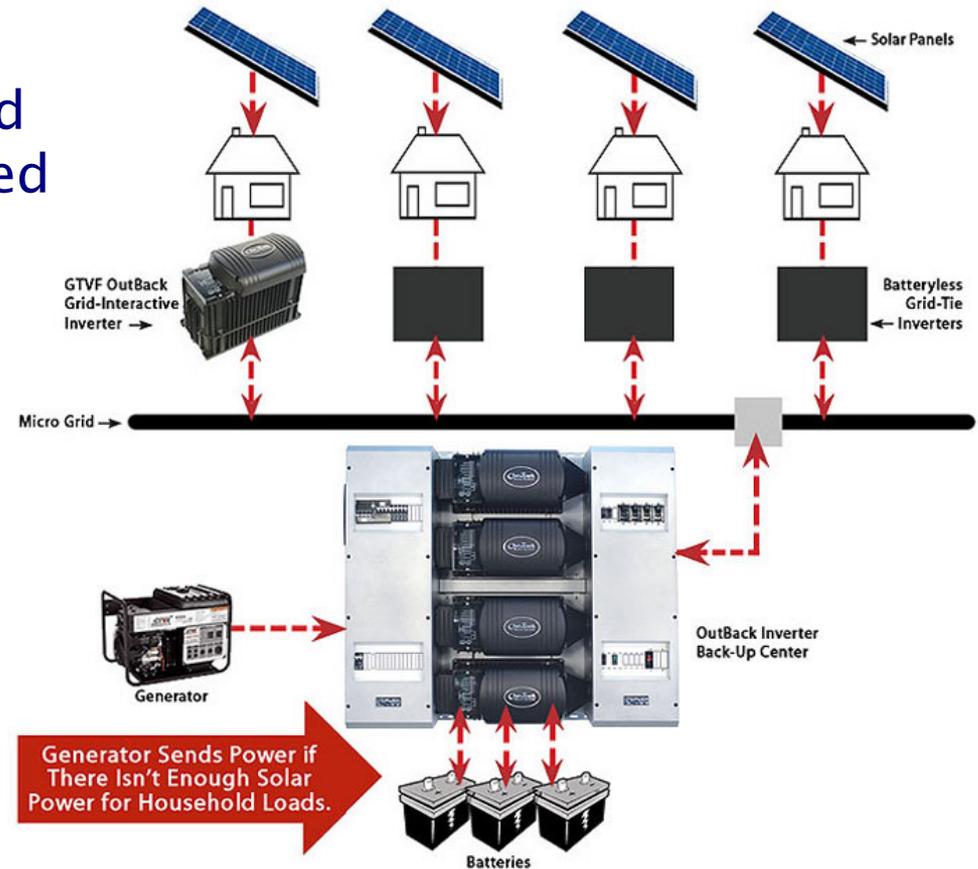
- **Prototype model checker for stochastic games**
 - integrated into PRISM model checker
 - using new explicit-state model checking engine
- **SMGs added to PRISM modelling language**
 - guarded command language, based on Reactive Modules
 - finite data types, parallel composition, proc. algebra op.s, ...
- **rPATL added to PRISM property specification language**
 - implemented value iteration based model checking
- **Available now:**
 - <http://www.prismmodelchecker.org/games/>

Case studies

- Evaluated on several case studies:
 - team formation protocol [CLIMA'11]
 - futures market investor model [McIver & Morgan]
 - collective decision making for sensor networks [TACAS'12]
 - energy management in microgrids [TACAS'12]

Energy management in microgrids

- Microgrid: proposed model for future energy markets
 - localised energy management
- Neighbourhoods use and store electricity generated from local sources
 - wind, solar, ...
- Needs: demand-side management
 - active management of demand by users
 - to avoid peaks



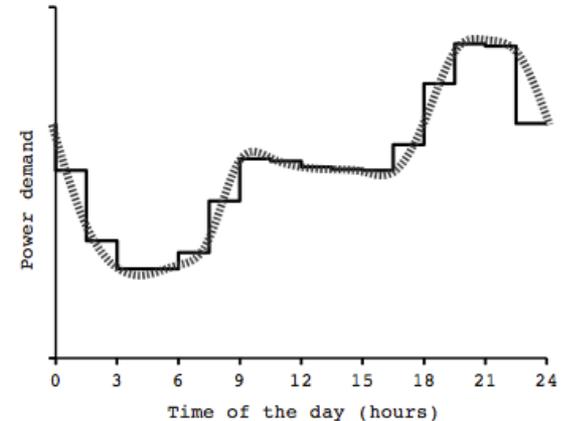
Microgrid demand-side management

- Demand-side management algorithm [Hildmann/Saffre'11]
 - N households, connected to a distribution manager
 - households submit loads for execution
 - load submission probability: daily demand curve
 - load duration: random, between 1 and D steps
 - execution cost/step = number of currently running loads
- Simple algorithm:
 - upon load generation, if cost is below an agreed limit c_{lim} , execute it, otherwise only execute with probability P_{start}
- Analysis of [Hildmann/Saffre'11]
 - define household value as $V = \text{loads_executing} / \text{execution_cost}$
 - simulation-based analysis shows reduction in peak demand and total energy cost reduced, with good expected value V
 - (if all households stick to algorithm)

Microgrid demand-side management

- **The model**

- SMG with N players (one per household)
- analyse 3-day period, using piecewise approximation of daily demand curve
- fix parameters $D=4$, $c_{lim}=1.5$
- add rewards structure for value V



- **Built/analysed models**

- for $N=2, \dots, 7$ households

- **Step 1: assume all households follow algorithm of [HS'11] (MDP)**

- obtain optimal value for P_{start}

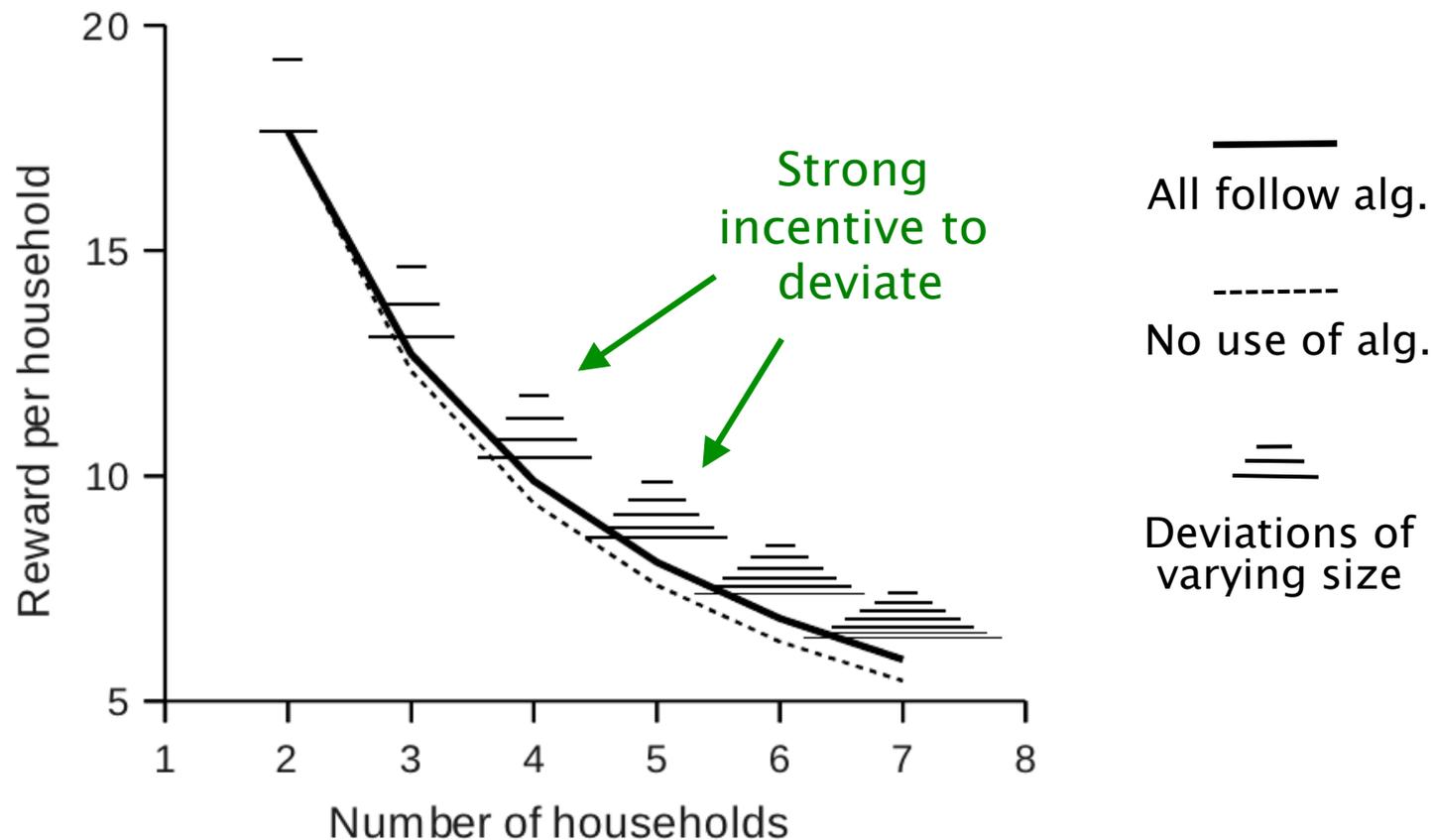
N	States	Transitions
5	743,904	2,145,120
6	2,384,369	7,260,756
7	6,241,312	19,678,246

- **Step 2: introduce competitive behaviour (SMG)**

- allow coalition C of households to deviate from algorithm

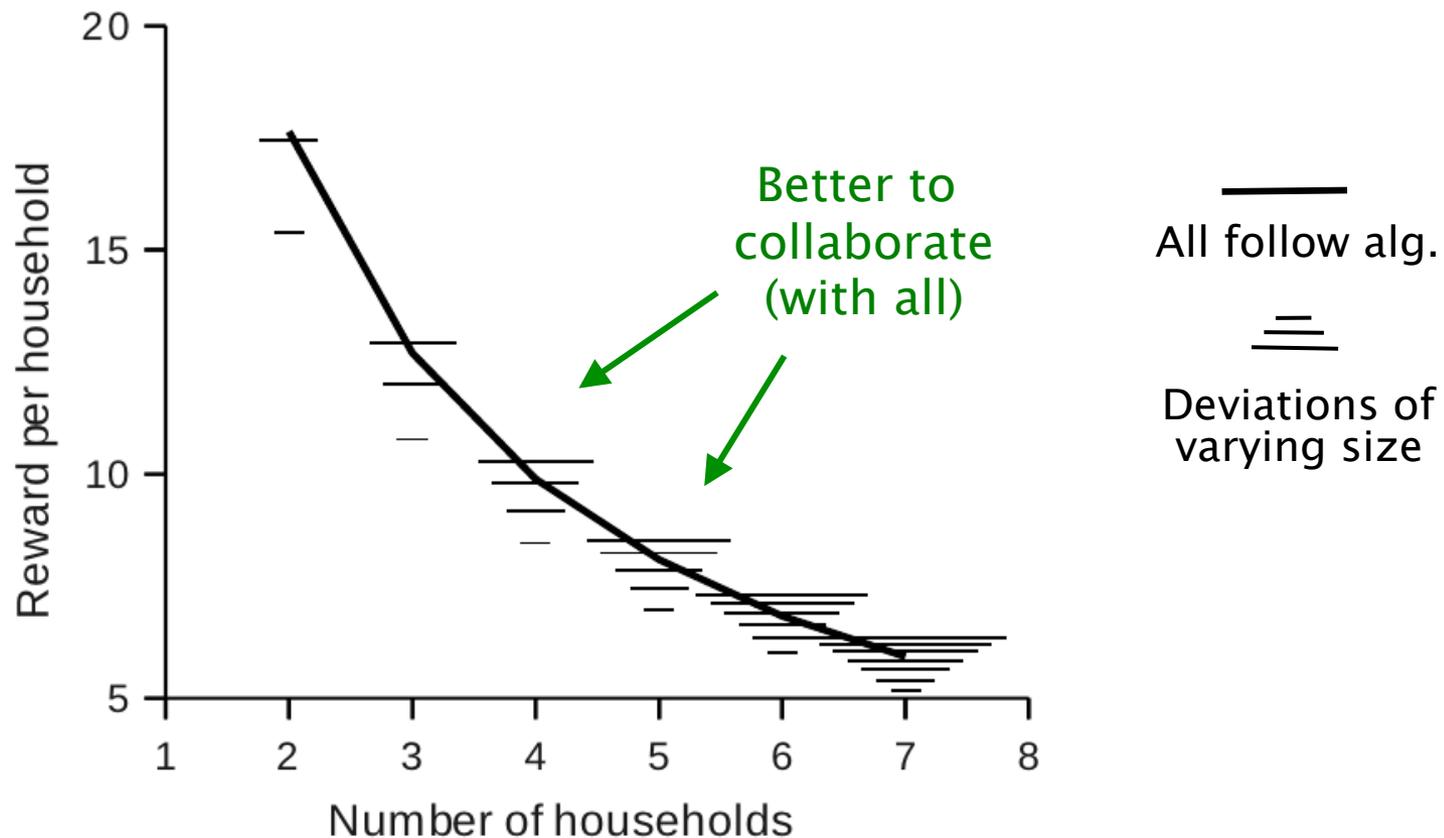
Results: Competitive behaviour

- Expected total value V per household
 - in rPATL: $\langle\langle C \rangle\rangle R^r c_{\max=?} [F^0 \text{ time}=\text{max time}] / |C|$
 - where r_c is combined rewards for coalition C



Results: Competitive behaviour

- Algorithm fix: simple punishment mechanism
 - distribution manager can cancel some loads exceeding C_{lim}



Conclusions

- **Conclusions**

- verification for stochastic systems with competitive behaviour
- modelled as stochastic multi-player games
- new temporal logic rPATL for property specification
- rPATL model checking algorithm based on num. fixed points
- prototype model checker PRISM-games
- case studies: energy management for microgrid

- **Future work**

- more realistic classes of strategy, e.g. partial information
- further objectives, e.g. multiple objectives, Nash equilibria, ...
- new application areas, security, randomised algorithms, ...