

# Automatic Verification of Competitive Stochastic Systems

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Based on TACAS'12 [FMSD'13], TACAS'13 and SR'13

# Automated quantitative verification

#### Quantitative verification

- of systems with stochastic behaviour, against temporal logic
- e.g. due to unreliability, uncertainty, randomisation, ...
- probability, costs/rewards, time,  $\dots$
- 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, ...

# Probabilistic models

- Discrete-time Markov chains (DTMCs)
  - discrete states + probability
  - for: randomisation, unreliable communication media, ...
- Continuous-time Markov chains (CTMCs)
  - discrete states + exponentially distributed delays
  - for: component failures, job arrivals, molecular reactions, ...
- Markov decision processes (MDPs)
  - probability + nondeterminism (e.g. for concurrency)
  - for: randomised distributed algorithms, security protocols, ...
- Probabilistic timed automata (PTAs)
  - probability, nondeterminism + real-time
  - for wireless comm. protocols, embedded control systems, ...

# Probabilistic model checking

- Property specifications based on temporal logic
  - PCTL, CSL, probabilistic LTL, PCTL\*, ...
- Simple examples:
  - $P_{\leq 0.01}$  [ F "crash" ] "the probability of a crash is at most 0.01"
  - $S_{>0.999}$  [ "up" ] "long-run probability of availability is >0.999"
- Usually focus on quantitative (numerical) properties:
  - P<sub>=?</sub> [ F "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
  - model checking: graph analysis + numerical solution + ...
  - or statistical model checking (sampling of executions, statistical tests or probability estimation)
- Probabilistic properties
  - $P_{max=?}$  [  $F^{\leq 10}$  "fail" ] "worst-case probability of a failure occurring within 10 seconds, for any possible scheduling of system components"
  - $P_{max=?}$  [  $G^{\leq 0.02}$  !"deploy" {"crash"}{max} ] "the maximum probability of an airbag failing to deploy within 0.02s, from any possible crash scenario"
- Reward-based properties (rewards = costs = prices)
  - R<sub>{"time"}=?</sub> [ F "end" ] "expected algorithm execution time"
  - $R_{\{"energy"\}max=?}$  [  $C^{\leq 7200}$  ] "worst-case expected energy consumption during the first 2 hours"

# The PRISM tool

- PRISM: Probabilistic symbolic model checker
  - developed at Birmingham/Oxford University, since 1999
  - free, open source (GPL), runs on all major OSs
- Support for:
  - discrete-/continuous-time Markov chains (D/CTMCs)
  - Markov decision processes (MDPs)
  - probabilistic timed automata (PTAs)
  - PCTL, CSL, LTL, PCTL\*, costs/rewards, ...
- Multiple efficient model checking engines
  - mostly symbolic (BDDs) (up to  $10^{10}$  states,  $10^7$ - $10^8$  on avg.)
  - widely used, 30,000 downloads
  - 100+ case studies,300+ papers
- See: <u>http://www.prismmodelchecker.org/</u>



# Modelling cooperation & competition

- Consider systems organised into communities
  - self-interested agents, goal driven
  - need to cooperate, e.g. in order to share bandwidth
  - possibly opposing goals, hence competititive behaviour
  - incentives to increase motivation and discourage selfishness
- Many typical scenarios
  - e.g. energy management, user-centric networks, or sensor network coordination
- Natural to adopt a game-theoretic view
  - widely used in computer science, economics, ...
  - here, distinctive focus on algorithms, automated verification
- <u>Research question</u>: can we <u>automatically verify</u> cooperative and competitive behaviour?

### 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):
  - $\Pi$  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
  - $-\Delta: S \times A \rightarrow Dist(S)$  is a (partial) transition probability function
  - $L: S \rightarrow 2^{AP}$  is a labelling with atomic propositions from AP
- Notation:
  - A(s) denotes available actions in state A



### Paths, strategies + probabilities

- A path is an (infinite) sequence of connected states in SMG
  - i.e.  $s_0a_0s_1a_1...$  such that  $a_i \in A(s_i)$  and  $\Delta(s_i,a_i)(s_{i+1}) > 0$  for all i
  - represents a system execution (i.e. one possible behaviour)
  - to reason formally, need a probability space over paths
- A strategy for player  $i \in \Pi$  resolves choices in  $S_i$  states
  - based on history of execution so far
  - − i.e. a function  $\sigma_i$  : (SA)\*S<sub>i</sub> → Dist(A)
  - $-\Sigma_i$  denotes the set of all strategies for player I
- A strategy profile is tuple  $\sigma = (\sigma_1, ..., \sigma_n)$  for n players
  - deterministic if  $\boldsymbol{\sigma}$  always gives a Dirac distribution
  - memoryless if  $\sigma(s_0a_0...s_k)$  depends only on  $s_k$
  - finite memory ...

#### Paths, strategies + probabilities...

#### For a strategy profile σ:

- the game's behaviour is fully probabilistic
- essentially an (infinite-state) Markov chain
- yields a probability measure  $Pr_s^{\sigma}$  over set of all paths  $Path_s$  from s

#### Allows us to reason about the probability of events

- under a specific strategy profile  $\boldsymbol{\sigma}$
- e.g. any ( $\omega$ -)regular property over states/actions
- Also allows us to define expectation of random variables
  - i.e. measurable functions  $X : Path_s \rightarrow \mathbb{R}_{\geq 0}$
  - $E_s^{\sigma}[X] = \int_{Path_s} X dPr_s^{\sigma}$
  - used to define expected costs/rewards...

### Rewards

- Rewards (or costs, prices)
  - real-valued quantities assigned to states (and/or transitions)
- Wide range of possible uses:
  - elapsed time, power 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  $\star \in \{\infty, c, 0\}$ , differing where T is not reached
  - reward is assumed to be infinite, cumulated sum, zero, resp.
  - $-\infty$ : 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  ${\bf R}$  from PRISM

#### • Example:

- $\langle \langle \{1,2\} \rangle \rangle P_{<0.01}$  [  $F^{\leq 10}$  error ]
- "players 1 and 2 have a strategy to ensure that the probability of an error occurring within 10 steps is less than 0.1, regardless of the strategies of other players"



#### rPATL syntax

• Syntax:

$$\begin{split} \varphi &::= \top \mid a \mid \neg \varphi \mid \varphi \land \varphi \mid \langle \langle C \rangle \rangle P_{\bowtie q}[\psi] \mid \langle \langle C \rangle \rangle R^{r}_{\bowtie x} \ [F^{\star}\varphi] \\ \psi &::= X \ \varphi \mid \varphi \ U^{\leq k} \ \varphi \mid F^{\leq k} \ \varphi \mid G^{\leq k} \ \varphi \end{split}$$

#### • where:

- a∈AP is an atomic proposition, C⊆Π is a coalition of players,  $\bowtie \in \{\le, <, >, \ge\}, q \in [0,1] \cap \mathbb{Q}, x \in \mathbb{Q}_{\ge 0}, k \in \mathbb{N} \cup \{\infty\}$ 

**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  $\psi$  being true satisfies  $\bowtie$  q, regardless of the strategies of other players"
- $\langle \langle C \rangle \rangle R^{r}_{\bowtie x} [F^{\star} \varphi]$ 
  - "players in coalition C have a strategy to ensure that the expected reward r to reach a  $\phi$ -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  $\Pi \backslash C$  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  $G_C$ :
  - $-\ \exists \sigma_1 {\in} \Sigma_1 \text{ such that } \forall \sigma_2 {\in} \Sigma_2 \text{ . } Pr_s^{\sigma_1,\sigma_2}(\psi) \bowtie q$
- Semantics for R operator defined similarly...

# Examples



 $\langle \langle \bigcirc \rangle \rangle P_{\geq \frac{1}{4}} [F \checkmark ]$ true in initial state

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

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

# Examples



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### Examples



 $\langle \langle \bigcirc \rangle \rangle P_{\geq \frac{1}{4}} [F \checkmark ]$ true in initial state

 $\langle \langle \bigcirc \rangle \rangle P_{\geq \frac{1}{3}} [F \checkmark]$ false in initial state

 $\langle \langle \bigcirc, \square \rangle \rangle P_{\geq \frac{1}{3}} [F \checkmark]$ true in initial state

### 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
  - $\langle \langle \text{sensorl} \rangle \rangle P_{<0.01} [ F (\neg \langle \langle \text{repairer} \rangle \rangle P_{\ge 0.95} [ F \text{"operational"} ] ) ]$

# Model checking rPATL

- Basic algorithm: as for any branching-time temporal logic
  - recursive descent of formula parse tree
  - compute  $Sat(\phi) = \{ s \in S \mid s \models \phi \}$  for each subformula  $\phi$
- Main task: checking P and R operators
  - reduction to solution of stochastic 2-player game  $G_C$
  - $\text{ e.g. } \langle \langle C \rangle \rangle P_{\geq q}[\psi] \ \Leftrightarrow \ \text{sup}_{\sigma_1 \in \Sigma_1} \text{ inf}_{\sigma_2 \in \Sigma_2} \text{ Pr}_s^{\sigma_1, \sigma_2}(\psi) \geq q$
  - complexity: NP  $\cap$  coNP (without any R[F<sup>0</sup>] operators)
  - compared to, e.g. P for Markov decision processes
  - complexity for full logic: NEXP  $\cap$  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  $\varphi$ ] : max/min reachability probabilities
  - compute  $\sup_{\sigma_1 \in \Sigma_1} \inf_{\sigma_2 \in \Sigma_2} \Pr_s^{\sigma_1, \sigma_2}(F \varphi)$  for all states s
  - deterministic memoryless strategies suffice
- Value is:
  - -1 if  $s \in Sat(\varphi)$ , 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
- Can also generate strategies

# Example



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

Player 1: ○, Player 2: ◆

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

### 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
- Strategy generation implemented
  - can generate strategies (memoryless, finite-memory for R[F<sup>0</sup>])
  - perform model checking under a strategy
- Available now [TACAS 2013]:
  - <u>http://www.prismmodelchecker.org/games/</u>



#### Case studies

- Applicable to strategic analysis of
  - distributed agreement protocols
  - reputation/virtual currency systems
- Evaluated on several case studies:
  - team formation protocol [CLIMA'11]
  - futures market investor model [Mclver & Morgan]
  - collective decision making for sensor networks [TACAS'12]
  - energy management in microgrids [TACAS'12]
  - user-centric networks [SR '13]

# 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,  $\dots$
- 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 probabilistic 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=loads\_executing/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,...,7 households
- Step 1: assume all households follow algorithm of [HS'11] (MDP)
  - obtain optimal value for  $\mathrm{P}_{\mathrm{start}}$





- 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<sup>0</sup> time=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 and strategy synthesis for stochastic systems with competitive behaviour
- modelled as stochastic multi-player games
- temporal logic rPATL for property specification
- rPATL /rPATL\* model checking algorithm based on numerical fixed points
- prototype tool PRISM-games
- case studies

#### Future work

- further objectives, e.g. multiple objectives
- correct-by-construction controller synthesis
- more realistic classes of strategy, e.g. partial information
- new application areas, security, randomised algorithms, ...