

We model stochastic systems through **Markov chains** and specify and evaluate their properties using probabilistic temporal logics and automata. In general, **logics** offer a clearer syntax and **automata** provide better performance in terms of computability. Therefore, it is important to define classes of logics and automata that have the **same expressive power** and can be used interchangeably. Here, we focus on two such formalisms known as **μ^p -calculus** and **p-automata**.

Background

μ^p -Calculus

The μ^p -calculus [1] is a probabilistic temporal logic. Formulas are built up from the combination of:

• Atomic propositions	$p, \neg p$
• Boolean connectives	\vee, \wedge
• Next operator	$O\varphi$
• Probabilistic quantification	$[\varphi]$
• Fixpoints	$\mu X. \varphi, \nu X. \varphi$

Using the fixpoint operators this logic can express **finite** and **infinite** iterations of properties:

Least fixpoint	μ	Finitely many iterations
Greatest fixpoint	ν	Infinitely many iterations

Formulas φ contained inside a probabilistic quantification are associated with a probability value in $[0,1]$. The operator $[\cdot]$ checks whether the value of the formula meets the **bound J** (of the form $\geq x$ or $> x$), and gets the value 1 or 0 accordingly. Therefore, top-level formulas are qualitative: either true or false.

When a μ^p -calculus formula is true on a Markov chain, we say that the **Markov chain satisfies the formula**.

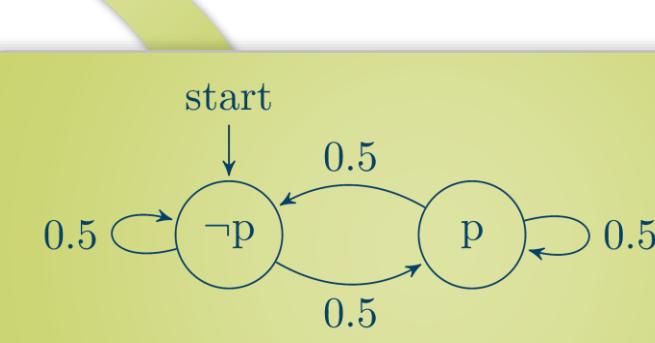
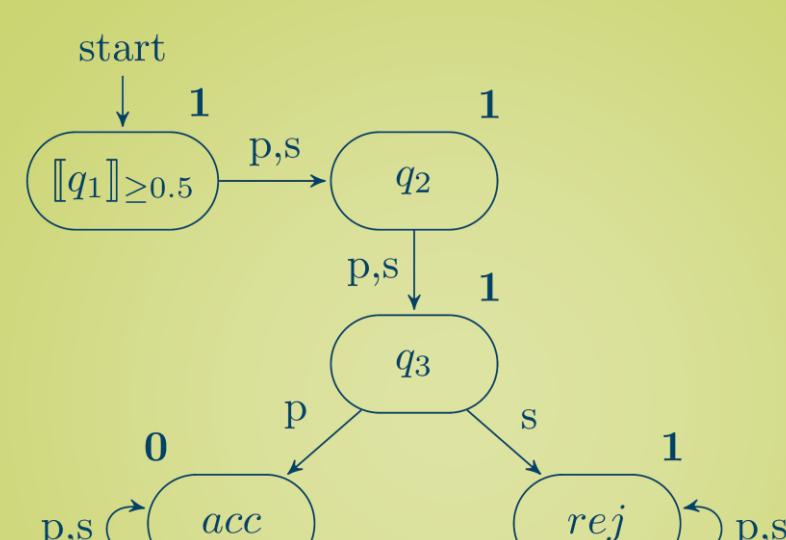


Fig. 1. Markov chain.

$[\Box p]_{\geq 0.5}$ The probability of reaching p in one step is ≥ 0.5
 $[\Box \Box p]_{\geq 0.5}$ The probability of reaching p in two steps is ≥ 0.5
 $[p \vee \Box p]_{\geq 0.5}$ The probability of p either now or in one step is ≥ 0.5

 Fig. 2. μ^p -Calculus formulas true on the Markov chain of Fig. 1.

 Fig. 3. A graph representing a p-automaton with: alphabet {p,s}, states {q1, q2, q3, acc, rej}, transitions=edges, initial condition $[\Box q_1]_{\geq 0.5}$.

Markov Chains

A Markov chain is a probabilistic transition system defined by the four components:

- 1) Set of **Locations**;
- 2) **Initial location**;
- 3) **Probability function**;
- 4) **Labelling function**.

The probability of moving from one location to each of its successors is a number in $[0,1]$. The sum of probabilities over all successors must be equal to 1.

p-Automata

A p-automaton [2] is an automaton that reads a Markov chain as **input** and decides whether to **accept** it or not. It is characterised by five components:

1. **States** are the elementary blocks and, to handle probabilities, may be enclosed in a probabilistic quantification $[\cdot]$.
2. **Alphabet** contains symbols that are read by the automaton, triggering a specific transition.
3. **Transitions** allow the automaton to move from one state to a Boolean (and/or) combination of them, depending on the symbol read.
4. **Initial condition** is a state, or a combination thereof, from which the automaton begins its computation.
5. **Acceptance** assigns a number to each state. Only states that are marked by an even number can be visited infinitely often.

Equivalence

μ^p -Calculus \rightarrow p-Automata

For every μ^p -calculus formula we can construct a p-automaton that accepts exactly those Markov chains that satisfy the formula [3].

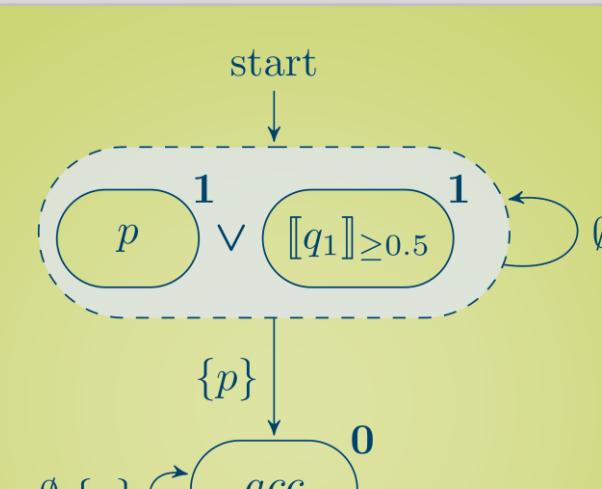
The components of the automaton resulting from the conversion are:

1. **States** originate from sub-formulas of the form: propositions, negated propositions, next, and quantified next; plus accepting and rejecting states.
2. **Alphabet** is the powerset of propositions appearing in the formula.
3. **Transitions** preserve the Boolean connectives (\vee, \wedge) and unfold the next operators into their nested sub-formulas.
4. **Initial condition** derives from the main formula without fixpoints.
5. **Acceptance** reflects the type of fixpoints that enclose the sub-formula/state ($\mu \leftrightarrow$ odd, $\nu \leftrightarrow$ even) and their potential nesting. Accepting and rejecting states are assigned numbers 0 and 1, respectively.

 $\mu X.(p \vee [\Box X]_{\geq 0.5})$

“Eventually reach a p within single steps whose probability is ≥ 0.5 ”

Fig. 4. μ^p -Calculus formula that either reads a p or performs a step and starts again. Since μ allows finitely many repetitions, a p must be reached eventually.


 Fig. 5. p-automaton that either reads a p or checks again. Since the topmost states can be visited only finitely many times, a p must be read eventually.

p-Automata \rightarrow μ^p -Calculus

For every p-automaton we can construct a μ^p -calculus formula satisfied in exactly those Markov chains accepted by the automaton [3].

The translation exploits the parallel between components of the automaton and elements of μ^p -calculus formulas:

- **Propositions** are taken from the alphabet.
- **Boolean connectives** match the and/or combinations of states defined by the transitions and initial condition.
- **Next** reflects the automaton's transitions.
- **Probabilistic quantification** is placed corresponding to bounded states of the automaton.
- **Fixpoints** are decided by looking at those states that are visited indefinitely and their acceptance number.

We have summarised the analogies that allow the **translation** from μ^p -calculus to p-automata and backwards. The mutual correspondence of the two languages implies their **equivalence** in expressive power; thus, lifting the well-known connection between logics and automata theory to a **probabilistic** scenario.

References

- [1] Castro P., Kilmurray C., and Piterman N., Tractable Probabilistic Temporal Logics, 2015
- [2] Chatterjee K. and Piterman N., Obligation Blackwell Games and p-Automata, 2017
- [3] Cauli C. and Piterman N., Equivalence of μ^p -Calculus and p-Automata, 2017

