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Contributions to
LTL and ω-Automata
for Model Checking

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The work summarized here is the result of collaborations with many people. As the reader shall soon discover, I have made arrangements so that those co-authors cannot be ignored while reading the document. Nonetheless, let me express some particular thanks to some colleagues from Brno (Jan Strejček, Tomáš Babiak, František Blahoudek) and Dresden (Joachim Klein and David Müller), who have inspired several recent developments, and with whom I enjoy working, even if I had less time to do so recently.
1 Introduction

This document is a synthesis of the work I have done since my Ph.D. Because of size constraints, it is written as a teaser, trying to lure the reader into reading the cited papers to get the details.\footnote{All these papers can be downloaded from \url{https://www.lrde.epita.fr/~adl/dl/adl.html}.}

The citations in this report have received a special treatment: references to papers I co-authored are shown in the margins inside such a blue frame, while third-party papers only appear in the final bibliography. In addition to separating my contributions from the existing literature, this presentation also shows that these are not precisely my contributions, but the results of collaborative work with several people.

Since this document is just baiting material, high-level explanations and examples have been preferred over formal definitions, and the reader is assumed to have some basic knowledge of linear-time temporal logic (LTL) and \(\omega\)-automata. Most of the work presented revolves around the automata-theoretic approach for LTL model checking (presented in the next section), and is implemented in the Spot library (presented in the following section).

1.1 Automata-theoretic approach to LTL model checking

The automata-theoretic approach to LTL model checking [Vardi, 2007] is a way to decide whether a model \(M\) satisfies an LTL formula \(\varphi\). The model \(M\) is first converted into an automaton (usually a Kripke structure) \(K_M\) whose language \(L(K_M)\) represents the set of all (infinite) behaviors of \(M\). The negation of the formula \(\varphi\) is converted into an automaton \(A_{\neg\varphi}\) whose language \(L(A_{\neg\varphi})\) captures the forbidden behaviors. With these objects, testing whether \(M\) satisfies \(\varphi\) amounts to checking the emptiness of the product \(K_M \otimes A_{\neg\varphi}\) of these two automata: if \(L(K_M \otimes A_{\neg\varphi}) = \emptyset\), then \(M \models \varphi\).

In LTL model checking, words are infinite, and \(K_M\) and \(A_{\neg\varphi}\) are \(\omega\)-automata. This approach can be declined in various ways depending on the nature of these two automata, and how we implement their product and its emptiness check.

In what follows, we will assume that \(K_M\) is always a Kripke structure, but we consider various natures for \(A_{\neg\varphi}\), different products, and different emptiness checks (Fig. 1.1).
1.2 Spot

Spot is a C++ library of model-checking algorithms that we first presented in 2004\textsuperscript{2}. It contains algorithms to perform the steps discussed in the previous section. It was purely a library until Spot 1.0, when we started distributing command-line tools for LTL manipulation\textsuperscript{3} and translation of LTL to some generalizations of Büchi automata.

The latest version\textsuperscript{4}, Spot 2.0, is a very large rewrite of the core of the library, in C++11, with a focus on supporting automata with arbitrary acceptance conditions, as we shall discuss in Chapter 4.

Spot now offers convenient command-line tools and Python binding to access most of its algorithms. A large part of my time over the last years has been spent trying to make these tools robust, useful, and accessible. Even if this sounds like engineering, building such a framework actually enables both research and teaching. Today, we have reached the point where several research teams are using Spot in part of their research or experiments\textsuperscript{5}, and we have started using it at Epita for teaching LTL, ω-automata and model checking.

Spot is a free software that can be downloaded from \url{https://spot.lrde.epita.fr/}. There is also a live installation of IPython/Jupyter (a web application for interactive programming [Pérez and Granger, 2007]) that can be used to experiment with Spot’s command-line tools or Python bindings without installing anything: see \url{http://spot-sandbox.lrde.epita.fr/}.

To emphasize that what we describe is not only implemented, but also easily reusable, this document is littered with examples using Spot from the command line. These examples are shown on a yellow background, as this paragraph.

Spot provides the following tools, that will be used in the examples:

- `randltl` generates random LTL/PSL\textsuperscript{6} formulas,
- `genltl` generates LTL formulas from scalable patterns,
- `ltlfilt` filters, converts, and transforms LTL/PSL formulas,
- `ltl2tgba` translates LTL/PSL formulas into generalized Büchi automata or deterministic parity automata,
- `ltl2tgta` translates LTL/PSL formulas into testing automata,
- `ltlcross` cross-compares LTL/PSL-to-automata translators to find bugs,
- `lctlgrind` mutates LTL/PSL formulas to help reproduce bugs on smaller ones,
- `dstar2tgba` converts the Rabin or Streett automata output by the `l12dstar` tool [Klein and Baier, 2006] into generalized Büchi automata,
- `randaut` generates random ω-automata,
- `autfilt` filters, converts, and transforms ω-automata,
- `ltldo` runs LTL/PSL formulas through other translators, providing uniform input and output interfaces.


\textsuperscript{5} Cf. Appendix A, p. 51.

\textsuperscript{6} PSL is a reference to the linear fragment of the Property Specification Language [Eisner and Fisman, 2006] which will only be mentioned in Section 7.2 (p. 40).
1.3 Overview of this document

Besides the introduction and conclusion, this habilitation thesis contains five core chapters. Each of them summarizes several related articles I co-authored.

In Chapter 2 we discuss how to translate $\varphi$ into $A\neg\varphi$, either as a Transition-based Generalized Büchi Automaton, a Büchi automaton, or a monitor. We summarize all the algorithms that have been implemented to make Spot’s $\text{lctl2tgba}$ one of the best off-the-shelf LTL translator tools.

In Chapter 3 we discuss various ways to test the emptiness of the above automata. This chapter was the topic of a Ph.D. I co-supervised [Renault, 2014], and introduces a new emptiness check procedure based on the union-find data structure.

In Chapter 4 we discuss a generic acceptance condition that we introduced in the Hanoi Omega-Automata (HOA) format in order to favor the interactions between $\omega$-automata tools. As these acceptance conditions supersede all conditions discussed previously, we present some of the acceptance transformations and adjustments we had to use to support algorithms over arbitrary acceptance in Spot.

In Chapter 5 we discuss a SAT-based technique for minimizing deterministic automata with arbitrary acceptance conditions.

In Chapter 6 we discuss several results related to stutter-invariant properties: how to test them, and how to implement the entire automata-theoretic approach using another type of automata called “testing automata”. The second part of this chapter was the topic of another Ph.D. I co-supervised [Ben Salem, 2014].

Some perspectives and omitted topics are gathered in Chapter 7. Finally, a list of selected uses of Spot is provided in Appendix A.
2

Translation improvements

2.1 Transition-based and generalized Büchi acceptance

Figure 2.1 shows four equivalent deterministic \( \omega \)-automata that are minimal for their respective acceptance conditions. These automata recognize infinite words over the alphabet \( \{\bar{a}b, \bar{a}b, \bar{a}b, \bar{a}b\} \) (i.e., all possible assignments of \( a \) and \( b \)). For conciseness, edges are labeled by Boolean formulas: they match all compatible assignments.

Small acceptance marks like \( \bullet \) or \( \circ \) are used to denote the membership of the states or transitions to different acceptance sets. In automata using state-based acceptance (Fig. 2.1(a)–(b)), these marks are on states. Automata with transition-based acceptance have them on transitions (Fig. 2.1(c)–(d)).

In an automaton with Büchi acceptance (Fig. 2.1(a) and (c)), a run is accepting if it visits infinitely many states or transitions with the mark \( \bullet \). Runs of automata with generalized Büchi acceptance (Fig. 2.1(b) and (d)) have to visit one \( \bullet \) and one \( \circ \) infinitely often. Generalized Büchi automata can use arbitrarily many acceptance marks; but in this particular example using more than 2 would not reduce the number of states.

We abbreviate Büchi automaton as BA, and prefix it with G or T to denote generalized or transition-based acceptance.

Any of these automaton types can be used as \( A_{\neg \varphi} \) in the model checking approach outlined in Fig. 1.1. However, the number of states of the product \( S = |K_M \otimes A_{\neg \varphi}| \) is such that \( S \leq |K_M| \cdot |A_{\neg \varphi}| \) so it seems better to pick the smallest automaton possible. Example 2.2 shows how to compute these sizes with a random \( K_M \) using Spot.

Clearly, BAs are a particular case of GBAs. Furthermore, any automaton with state-based acceptance can easily be converted into an automaton with transition-based acceptance without changing its transition structure: it suffices to push the marks of each state to all their outgoing transitions (Fig. 2.3). With this interpretation, we can consider automata with state-based acceptance as particular cases of automata with transition-based acceptance.

Example 2.2: Commands for computing the size of \( K_M \otimes A_{\neg \varphi} \) when \( K_M \) is a random 100-state automaton with an edge density of 0.05 (each state has 5 successors on average), and \( K_M \) is either a TGBA (Fig. 2.1(d)) or a BA (Fig. 2.1(a)).

\[
\begin{align*}
\texttt{ltl2tgba } 'GFA \& GFb' & \rightarrow \texttt{tgba} \\
\texttt{ltl2tgba } -B 'GFA \& GFb' & \rightarrow \texttt{ba} \\
\texttt{randaut } -Q100 \ -e.05 \ a \ b \ >k \\
\texttt{autfilt } --product=k \ \texttt{tgba ba} \\
& > --stats='%F: %s st, %e ed' \\
\texttt{tgba: 100 st, 589 ed} \\
\texttt{ba: 284 st, 1681 ed}
\end{align*}
\]

Figure 2.1: Minimal deterministic automata recognizing the LTL formula \( GFa \land GFb \). These automata recognize infinite words over the alphabet \( \{\bar{a}b, \bar{a}b, \bar{a}b, \bar{a}b\} \) (i.e., all possible assignments of \( a \) and \( b \)). For conciseness, edges are labeled by Boolean formulas: they match all compatible assignments.

Figure 2.3: A TGBA interpretation of the state-based GBA of Fig. 2.1(b).
Finally, a TGBA with \( n \) marks and \( s \) states can be converted into a TBA with at most \( n \cdot s \) states or into a BA with at most \( (n + 1) \cdot s \) states, using a degeneralization procedure we will discuss later. This shows that all these automata are as expressive, but that TGBAs can be more concise. For this reason, most of our work was focused on transition-based acceptance.\(^1\)

2.2 Pros and cons of transition-based acceptance

From an implementation point of view, using transition-based acceptance requires more memory, because acceptance marks have to be stored on transitions, which are more numerous than states\(^2\). This is true as long as the automaton is explicitly stored in memory, and this cost cannot always be compensated by the smaller size of automata with transition-based acceptance. However, when implementing an explicit model checker using the automata-theoretic approach of Fig. 1.1, one should keep in mind that the product of \( K_M \otimes A_{\neg \varphi} \) is computed on the fly, as needed by the emptiness-check procedure, and that its transitions need not be stored: only the states of the product have to be kept in order to remember which ones were already visited. In that context, the only memory cost of using transition-based acceptance marks is in the representation of \( A_{\neg \varphi} \), which is negligible compared to the size of \( K_M \otimes A_{\neg \varphi} \).

Using transition-based acceptance often leads to simpler or more natural algorithms. This observation has been made several times in the past: for instance it is discussed by Michel [1984], Kurshan [1987], Le Saëc and Litovsky [1994], Giannakopoulou and Lerda [2002], and Varghese [2014]. After having taken several algorithms originally introduced in a state-based setting and implemented them in a transition-based framework, I can only second these observations.

However, we should keep in mind that there is a large subset of properties that do not benefit from transition-based Büchi acceptance in any way. (Inherently) Weak automata are automata in which any strongly connected component (SCC) contains only accepting cycles, or only non-accepting cycles. In these automata, the acceptance condition could be expressed directly in terms of SCCs rather than states or transitions: an infinite run is accepting if it ultimately stays in one of the accepting SCCs. There is also nothing to be gained by using generalized Büchi acceptance in weak automata, or by trying to represent the same properties using non-weak Büchi automata. This class of automata represents properties from the persistence class defined by Manna and Pnueli [1990] (Fig. 2.4), and can be associated with a syntactic fragment of LTL [Černá and Pelánek, 2003].

2.3 Benchmark of BA translators

Before we discuss how we translate LTL into TGBA or BA, let us make a quick review of existing tools. Most off-the-shelf translators focus on producing (state-based) Büchi automata, because that is what is still commonly used.

\(^{\dagger}\) Spot actually supports only transition-based acceptance. When it outputs automata with state-based acceptance, it is just pretending: internally those automata are stored using transition-based marks, as in Figure 2.3.

\(^{\dagger}\) Except in some degenerate cases, we work with automata that have at least one outgoing transition per state.

---

Example 2.5: Classification of the 55 LTL formulas of Dwyer et al. [1998] into their most precise class of the temporal hierarchy. Of these formulas, only 14 (12 recurrence and 2 reactivity properties) may not be represented by weak Büchi automata.
We consider the following translators:

<table>
<thead>
<tr>
<th>Translator</th>
<th>Version</th>
<th>References</th>
</tr>
</thead>
<tbody>
<tr>
<td>spin</td>
<td>6.4.5</td>
<td>[Holzmann, 2003]</td>
</tr>
<tr>
<td>lt12ba</td>
<td>1.2b1</td>
<td>[Gastin and Oddoux, 2001]</td>
</tr>
<tr>
<td>modella</td>
<td>1.5.93</td>
<td>[Sebastiani and Tonetta, 2003]</td>
</tr>
<tr>
<td>trans</td>
<td>Mar.2013</td>
<td>[Mochizuki et al., 2014]</td>
</tr>
<tr>
<td>lt13ba</td>
<td>1.1.3</td>
<td>[Babiak et al., 2012]</td>
</tr>
<tr>
<td>lt12tgba</td>
<td>Spot 2.3</td>
<td>[Duret-Lutz, 2014]</td>
</tr>
</tbody>
</table>

lt12ba was created as a drop-in replacement for the translator of spin, and reuses part of its source. modella’s motto was that larger automata are OK as long as they are more deterministic, because improved determinism should benefit the product performed by a model checker. lt13ba is a drop-in replacement for lt12ba where the core of the translation has been rewritten and some optimization were added. In particular it tries to favor more deterministic automata by default (this can be disabled with option \(-M0\). lt12tgba is our own translator: by default it favors small automata over deterministic ones, but this can be changed with option \(-D\). Finally, trans is a more recent translator that, unlike lt12ba, lt13ba and lt12tgba, does not use transition-based acceptance internally.

Table 2.7 compares these tools on the production of BAs. For a given formula \(\varphi\) and a tool \(t\) we compute a BA \(A_\varphi^t\) and count its number of states, non-deterministic states, edges, and transitions. The number of transitions is obtained by counting each edge for the number of compatible assignments of all atomic propositions. To measure the behavior of \(A_\varphi^t\) in a model checking context, we compute 100 products \(M_i \otimes A_\varphi^t\), using 100 randomly generated automata \((M_i)_{1 \leq i \leq 100}\) (the same automata are used for all tools), and retain the median size of these products. For each tool, the table sums all the sizes over 178 formulas taken from the literature.

Spot’s lt12tgba produces Büchi automata that are noticeably smaller and more deterministic than those produced by other tools, but the additional simplifications it performs have a cost. It takes nearly two seconds to translate all formulas, and that is three times the time needed by lt13ba.

<table>
<thead>
<tr>
<th>#</th>
<th>ndet</th>
<th>time (s)</th>
<th>states</th>
<th>ndstates</th>
<th>edges</th>
<th>trans</th>
<th>states</th>
<th>product size</th>
</tr>
</thead>
<tbody>
<tr>
<td>spin</td>
<td>167</td>
<td>162</td>
<td>220.66</td>
<td>1440</td>
<td>1236</td>
<td>7238</td>
<td>46033</td>
<td>259313</td>
</tr>
<tr>
<td>lt12ba</td>
<td>178</td>
<td>169</td>
<td>0.32</td>
<td>1000</td>
<td>801</td>
<td>3338</td>
<td>29974</td>
<td>190898</td>
</tr>
<tr>
<td>modella -r12 -g -e</td>
<td>178</td>
<td>109</td>
<td>18.47</td>
<td>1244</td>
<td>577</td>
<td>4116</td>
<td>23474</td>
<td>210494</td>
</tr>
<tr>
<td>trans</td>
<td>178</td>
<td>119</td>
<td>0.46</td>
<td>957</td>
<td>398</td>
<td>4116</td>
<td>23474</td>
<td>172246</td>
</tr>
<tr>
<td>lt13ba -M0</td>
<td>178</td>
<td>167</td>
<td>0.72</td>
<td>794</td>
<td>591</td>
<td>2193</td>
<td>21170</td>
<td>150776</td>
</tr>
<tr>
<td>lt13ba</td>
<td>178</td>
<td>115</td>
<td>0.72</td>
<td>829</td>
<td>307</td>
<td>2389</td>
<td>14322</td>
<td>155220</td>
</tr>
<tr>
<td>lt12tgba -s</td>
<td>178</td>
<td>49</td>
<td>1.85</td>
<td>666</td>
<td>102</td>
<td>1643</td>
<td>10346</td>
<td>129419</td>
</tr>
<tr>
<td>lt12tgba -s -D</td>
<td>178</td>
<td>44</td>
<td>1.87</td>
<td>671</td>
<td>96</td>
<td>1667</td>
<td>10456</td>
<td>129804</td>
</tr>
</tbody>
</table>
2.4 Translating LTL formulas into Büchi automata: an overview

The operations chained in Spot to translate an LTL formula into a Büchi automaton can be grouped into four steps:

1. Simplify the LTL formula syntactically. E.g., rewrite $FFa$ (a 3-state automaton) into $Fa$ (2 states).
2. Translate the simplified formula into a TGBA.
3. Post-process the resulting TGBA, e.g., by pruning useless SCCs, or running various simulation-based reductions or minimizations.
4. If desired (and still needed after the previous post-processing) degeneralize the TGBA into a Büchi automaton.

These steps are summarized in the next four sections, but more details can be found in our IJCCBS paper\(^5\).

2.5 LTL simplifications

Spot implements several types of rewritings:

- **Trivial identities** are applied at any time during the construction of a formula (e.g., while they are parsed). These are all based on idempotence of some operators (e.g., $FFa \equiv Fa$), or neutral/absorvent operands (e.g., $X \bot \equiv \bot$, $f \land \bot \equiv \bot$, $f \land \top \equiv \top$, etc.).
- **Basic rewritings** are unconditional rewriting rules, like moving the $X$ operators to the front (as in $GXf \equiv XGf$).
- **Eventual and universal rewritings** apply only when some subformulas are purely universal or pure eventualties [Etessami and Holzmann, 2000], or are what Babiak et al. [2012] have called alternating formulas. As an example $FGFa$ can be rewritten as $GFa$ because the latter is a pure eventuality.
- **Implication-based rewritings** apply only in cases where one subformula can be shown to imply another subformula. For instance, under the hypothesis that $f \to g$, we have $f U g \equiv g$. There are two ways to detect such implications: they can be approximated syntactically [Somenzi and Bloem, 2000], or decided exactly using automata-based language containment checks [Tauriainen, 2006].

Not all the rewriting rules found in the literature are necessarily good to apply\(^6\): some were written with the goal of reducing the size or the depth of the formula, but these metrics do not always correlate with the size of the automaton. For instance the rewriting of $F(\varphi \land GF\psi)$ into $(F\varphi) \land (GF\psi)$ suggested by Somenzi and Bloem [2000] is actually harmful in our context: our translation works better with the first formula, where it is clear that $GF\psi$ does not need to be checked before $\varphi$ holds.

2.6 Translating LTL to TGBA

The core translation algorithm implemented is based on an original algorithm by Couvreur [1999] that uses binary decision diagrams

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\(^6\) An exhaustive list of all simplification rules implemented in Spot can be found in https://spot.lrde.epita.fr/tl.pdf
(BDDs) to represent the outgoing transition of each individual state. With this representation two states that have exactly the same successors have the same BDD representation and can be merged. This algorithm has been improved in a couple of ways:

- It uses the BDD representation to improve the determinism of the resulting automaton. Briefly, this comes from the fact that a formula $a \lor (b \land X\varphi)$ where the choice between $a$ and $b$ is not exclusive (both could hold) can also be interpreted as the equivalent $a \lor (\bar{a} \land b \land X\varphi)$ where the choice between $a$ and $\bar{ab}$ is exclusive.

- When it is syntactically obvious that the property translated is persistent (Fig. 2.4), it forces the output automaton to be weak, as suggested by Černá and Pelánek [2003].

Our IJCCBS paper describes a few other minor improvements (faster translation for subformulas of G, early simplification of the number of acceptance sets) that are harder to present without going into details.

2.7 Simplifying TGBAs

The TGBAs resulting from the above translation can often be simplified. We have implemented three types of simplifications. The first three of these steps are described in detail in our Spin’13 paper.¹

1. SCC-based simplifications use the fact that accepting runs of Büchi automata will ultimately end in one (accepting) SCC of the automaton. States that cannot reach an accepting SCC are useless and can be removed. SCCs that are not accepting do not need acceptance marks.

2. Acceptance mark simplifications: if the set of transitions with $\bullet$ includes the set of transitions with $\bullet$, then the mark $\bullet$ is superfluous and can be removed from the transition structure as well as from the acceptance condition. This inclusion check can be refined on an SCC-per-SCC basis.

3. Simulation-based relations can be used to merge states and remove transitions. Spot implements direct and reverse simulations; and those work with arbitrary acceptance conditions. The implementation of this reduction uses a BDD-based representation of a “signature” of each state, that (as in the case of the main translation), automatically prunes redundant transitions, and improves the determinism of the resulting automaton.

4. Weak and deterministic Büchi automata (WDBA) can be minimized efficiently by an algorithm due to Löding [2001]. Even if the result of the core translation does not give a WDBA, any obligation property (cf. Fig. 2.4) can be represented by a WDBA and Dax et al. [2007] show how to obtain it, if it exists.

All these optimizations can be applied to any automaton using the autfilt tool (Ex. 2.10).

The use of WDBA-minimization is probably the main reason for the success of ltl2tgba in Table 2.7. In many cases, the minimal WDBA obtained is smaller than the non-deterministic automaton

```python
import spot
f = spot.formula('GFa->GFb')
aut = spot.ltl_to_tgba_fm(f, spot._bdd_dict)
print(aut.to_str('HOA'))
```

Example 2.9: The core LTL translation algorithm cannot be easily accessed from the command-line: the ltl2tgba tool will always perform some amount of formula and automaton simplification, even in its lowest settings. Experiments with this low-level algorithm can be done in C++ or Python (above) by calling the ltl_to_tgba_fm() function.

Example 2.10: Spin translates XF$_a$ into a 4-state BA. Simplifying that automaton with autfilt produces a 3-state BA.


produced by the core translation algorithm. The other cases raise an interesting question (cf. Fig. 2.11): should a small non-deterministic automaton be preferred to a larger deterministic one? ltl2tgba and autfilt leave that decision to the user: the -D option favors deterministic automata even when larger.

2.8 From TGBA to BA

Because several model checking tools only work with (state-based) Büchi automata, Spot has a rather evolved degeneralization routine.

Transforming the TGBA of Figure 2.1(d) (p. 9) into the BA of Figure 2.1(a) can be done with a classical procedure that duplicates the original TGBA \( n + 1 \) times where \( n \) is the number of acceptance marks. Improvements to this text-book procedure were discussed by Gastin and Oddoux [2001], and this can be used to build the BA of Figure 2.12(b).

At SPIN'13 we presented two additional optimizations that use information about SCCs in the TGBA to avoid creating some useless states in the resulting BA. This optimized degeneralization procedure builds the BA of Figure 2.12(c).

![Input TGBA](image1.png)

(a) input TGBA

![Equivalent BA obtained with classical degeneralization](image2.png)

(b) Equivalent BA obtained with classical degeneralization

![Equivalent BA produced by SCC-aware degeneralization](image3.png)

(c) Equivalent BA produced by SCC-aware degeneralization

2.9 Optimizing BA for Spin

The previous result was a collaboration with the authors of ltl3ba, and we continued this collaboration by exploring how the shape of the BA could influence the performance of the spin model checker. This line of work produced two articles, at Spin’14 and Spin’15.

In the first one, we compared several automata produced by off-the-shelf translators, as done for example in Fig. 2.13, and observed that using them with Spin to model-check an actual model from the BEEM benchmark [Pelánek, 2007] gives some counterintuitive results. As can be seen in Table 2.14 the size of the property automaton does not correlate with the run time of Spin. Automata \( D_2 \) and \( D_4 \) have the same size (for all metrics used in this table), but show a two-fold difference in model-checking performance. Automaton \( D_7 \) is slightly worse than \( D_6 \) despite having fewer edges.

Considering the problem from the point of view of the emptiness
check algorithm used by Spin, which is based on two nested depth-first searches [Holzmann et al., 1996], we realized that the location of accepting states in the automaton is important.

We concluded that when the product $K_M \otimes A_{\neg \varphi}$ is expected to be empty (\(\varphi\) should hold), the best automaton $A_{\neg \varphi}$ for Spin should have accepting states that are hard to reach from the initial state as it will lessen the chance to trigger a nested DFS. On the contrary, if the product $K_M \otimes A_{\neg \varphi}$ contains an accepting cycle, Spin can find it faster if the accepting states of $A_{\neg \varphi}$ are easy to reach from the initial state and the accepting cycles are short. It turns out that the degeneralization procedure discussed in the previous section can be adjusted for both cases.

In a second paper\(^{11}\), we remarked that if an automaton $A_{\neg \varphi}$ is going to be used to verify a model $M$ by constructing the product

<table>
<thead>
<tr>
<th>automaton size</th>
<th>states</th>
<th>ndst</th>
<th>edges</th>
<th>trans</th>
<th>statistics from Spin’s execution</th>
<th>stored states</th>
<th>visited trans</th>
<th>time</th>
</tr>
</thead>
<tbody>
<tr>
<td>$D_1$ Spin</td>
<td>3</td>
<td>2</td>
<td>6</td>
<td>12</td>
<td></td>
<td>15777846</td>
<td>7680k</td>
<td>6.04s</td>
</tr>
<tr>
<td>$D_2$ LTL2BA</td>
<td>3</td>
<td>3</td>
<td>6</td>
<td>12</td>
<td></td>
<td>1577440</td>
<td>7684k</td>
<td>5.95s</td>
</tr>
<tr>
<td>$D_3$ MoDeLLa</td>
<td>5</td>
<td>2</td>
<td>8</td>
<td>18</td>
<td></td>
<td>1580893</td>
<td>7670k</td>
<td>6.13s</td>
</tr>
<tr>
<td>$D_4$ LTL3BA</td>
<td>3</td>
<td>3</td>
<td>6</td>
<td>12</td>
<td></td>
<td>2299250</td>
<td>15583k</td>
<td>12.10s</td>
</tr>
<tr>
<td>$D_5$ LTL3BA (det)</td>
<td>4</td>
<td>1</td>
<td>7</td>
<td>14</td>
<td></td>
<td>2297625</td>
<td>15561k</td>
<td>12.00s</td>
</tr>
<tr>
<td>$D_6$ LTL2TGBA</td>
<td>3</td>
<td>1</td>
<td>6</td>
<td>9</td>
<td></td>
<td>848641</td>
<td>2853k</td>
<td>2.26s</td>
</tr>
<tr>
<td>$D_7$ LTL2TGBA (no jmp)</td>
<td>3</td>
<td>1</td>
<td>5</td>
<td>9</td>
<td></td>
<td>852094</td>
<td>2863k</td>
<td>2.34s</td>
</tr>
<tr>
<td>$D_8$</td>
<td>3</td>
<td>1</td>
<td>6</td>
<td>9</td>
<td></td>
<td>848641</td>
<td>2853k</td>
<td>2.43s</td>
</tr>
<tr>
<td>$D_9$</td>
<td>3</td>
<td>3</td>
<td>6</td>
<td>11</td>
<td></td>
<td>852094</td>
<td>2878k</td>
<td>2.43s</td>
</tr>
<tr>
<td>$D_{10}$</td>
<td>3</td>
<td>1</td>
<td>7</td>
<td>10</td>
<td></td>
<td>1575844</td>
<td>7658k</td>
<td>7.38s</td>
</tr>
<tr>
<td>$D_{11}$</td>
<td>3</td>
<td>1</td>
<td>6</td>
<td>10</td>
<td></td>
<td>1577440</td>
<td>7657k</td>
<td>7.07s</td>
</tr>
<tr>
<td>$D_{12}$</td>
<td>3</td>
<td>1</td>
<td>6</td>
<td>10</td>
<td></td>
<td>2297625</td>
<td>15561k</td>
<td>12.30s</td>
</tr>
</tbody>
</table>

$K_M \otimes A_{\neg \varphi}$, then $A_{\neg \varphi}$ can be simplified by taking into account some knowledge of $M$. Typically we can use the fact that some of the atomic propositions used in $\varphi$ are mutually exclusive in $M$: for instance $x = 5$ and $x > 10$ cannot both hold at once, and a process cannot be in two different locations at the same time.

Using such a priori knowledge about the model, we can simplify the automaton, as illustrated by Fig. 2.15. The simplifications performed are two-fold.

First, the automaton can be refined by removing or merging useless transitions: for instance under the assumption that two propositions among $\{a, b, c\}$ cannot both hold at once, the behaviors captured by the top-left state of automaton $B_1$ are included in those captured by its top-right state, allowing the former state to be removed. This type of simplification entailed an average 28% reduction of the number of transitions visited by Spin on the empty products of our benchmark.

A second simplification is to shorten the Boolean formulas which label the transitions in order to limit the number of evaluations of atomic propositions performed by the model checker: for instance under the same assumptions, $\overline{bc}$ can be rewritten $c$, testing a single atomic proposition instead of two. This type of subtle change improved the run time of Spin by 3.5%.

### 2.10 Building unambiguous automata

An automaton is unambiguous if any word can be recognized by at most one accepting run. This can be understood as a weaker notion than determinism: in a deterministic automaton any word should be recognized by at most one run of the automaton (that could be a rejecting run). Unambiguous Büchi automata are as expressive as Büchi automata [Arnold, 1984, Theorem IV-1], and can be exponentially more succinct than deterministic BA [Bousquet and Löding, 2010, Remark 2].

Probabilistic model checking of an LTL property $\varphi$ against a Markov chain $M$ is classically done using an $\omega$-automaton $A_{\varphi}$ that is deterministic in order to guarantee that $M \otimes A_{\varphi}$ is a Markov chain [Baier and Katoen, 2008, Section 10.3]. However, unambiguous automata are an attractive alternative, since the probabilities really only have to be computed along accepting runs [Baier et al., 2016b].

As seen in Table 2.7, Spot is already good at producing small automata that are often deterministic (hence unambiguous). However, it also implements a variant of its translation algorithm that always outputs unambiguous TGBAs. The technique extends an idea described by Benedikt et al. [2013] to all operators supported by Spot.

This idea can be quickly explained as follows. When translating the LTL formula $a \mathcal{U} \beta$ as an automaton, this formula is normally interpreted as the equivalent formula $\beta \lor (a \land X(a \mathcal{U} \beta))$, meaning that a state recognizing the language $L(a \mathcal{U} \beta)$ will usually have two outgoing transitions for $L(\beta)$ and $L(a \land X(a \mathcal{U} \beta))$. To obtain an
unambiguous automaton, it suffices to make these two later languages disjoint. So in practice $\emptyset \cup \{ a \land \neg a \}$ would now be seen as $\beta \lor (a \land \neg b \land X(a \land b))$, therefore representing two outgoing transitions for the languages $\mathcal{L}(\beta)$ and $\mathcal{L}(a \land \neg b \land X(a \land b))$. This idea has to be generalized to all operators that involve a non-deterministic choice.

Figure 2.16 shows an ambiguous BA resulting from the translation of $GFa \rightarrow GFb \equiv FG \neg a \lor GFb$. One can readily see that the sub-automaton consisting of the states $\{0, 1\}$ recognizes $\mathcal{L}(GFb)$ while the sub-automaton $\{0, 2\}$ recognizes $\mathcal{L}(FG \neg a)$. Those two languages both contain the word $ab; ab; ab; \ldots$, and the $T$-labeled self-loop on state $0$ makes it possible to build an infinite number of accepting runs over this word in any of these two sub-automata.

In the unambiguous BA of Figure 2.17, the sub-automaton $\{0, 1\}$ still recognizes $\mathcal{L}(GFb)$, but the sub-automaton $\{0, 2, 3, 4\}$ now recognizes $\mathcal{L}(FG \neg a \land \neg GFb)$ which is disjoint. The word $ab; ab; ab; \ldots$ is now accepted by a unique run that goes through states $0$ and $1$.

Baier et al. [2016b] used unambiguous Büchi automata generated by Spot in their experiments, showing that they were a viable alternative to probabilistic model checking using deterministic Rabin automata. To our knowledge, the only other existing tool that can build unambiguous automata from LTL is Tulip [Lenhardt, 2013] However its website is not responsive anymore, and it was not clear whether its built-in LTL translator as a standalone tool was possible. Therefore we have never performed any tool comparison between unambiguous automata produced by Spot and those of any other tool.

2.11 Building monitors

Let us define monitors as a subclass of generalized Büchi automata where no acceptance sets are used, i.e., all runs are accepting. Since the only way to reject words is to use automata that are incomplete, monitors can be used to detect “bad prefixes”: the expressive power of monitors corresponds to the Safety class of Figure 2.4 (p. 10).

In the context of online runtime verification, a monitor for a correctness property $\varphi$ can be run alongside a live system to monitor its events, and signal any execution whose current (finite) prefix cannot be extended into an infinite word that is correct for $\varphi$.

D’Amorim and Roşu [2005] and Tabakov and Vardi [2010] explore this usage, and describe algorithms to construct monitors (and deterministic monitors) that reject the bad prefixes of any LTL formula $\varphi$. Once a TGBA for $\varphi$ has been obtained, it suffices to remove all useless SCCs, then remove all acceptance sets, and then optionally determinize and minimize the result. Even if $\varphi$ is not a safety property this procedure will build a monitor recognizing a superset of $\mathcal{L}(\varphi)$, effectively capturing the monitorable parts of $\varphi$.

These transformations are implemented in Spot; `ltl2tgba` can generate monitors with option `-M`, and deterministic monitors with options `-M -D`, as illustrated by Figures 2.18 and 2.19.
3  

Emptiness checks

This chapter was the subject of E. Renault’s Ph.D thesis [2014] which I co-supervised with F. Kordon and D. Poitrenaud.

3.1 Introduction

If an ω-automaton accepts a word, it necessarily contains an accepting cycle (i.e., a cycle visiting all acceptance marks in the case of generalized Büchi) that is reachable from the initial state. Emptiness check algorithms search for the existence of such a cycle.

Automata with generalized Büchi acceptance can be tested for emptiness by a slight modification of SCC-enumeration algorithms [Couvreur, 1999, Geldenhuys and Valmari, 2005]. Büchi automata can be checked using the same algorithms, but their simpler acceptance can also be verified using two nested depth-first search (NDFS) [Holzmann et al., 1996] with a smaller memory footprint. When the property to verify is more concisely expressed using generalized Büchi acceptance, it is preferable to use SCC-based algorithms over degeneralizing the automaton to use an NDFS.

3.2 Strength-based decompositions

Some subclasses of automata enable more efficient emptiness checks. In particular, weak automata (already discussed page 10, along with Fig. 2.4) can be tested for emptiness using a simple DFS [Černá and Pelánek, 2003]. Terminal automata form a subclass of weak automata in which accepting SCCs are complete. When the model M has no deadlock (each state has at least one successor) and the property automaton $A_{¬ϕ}$ is known to be terminal, the emptiness check of the product becomes a simple reachability problem: detecting cycles is not even necessary.

Based on these notions, we proposed to split an arbitrary automaton $A_{¬ϕ}$ into possibly three automata $A_S$, $A_W$, and $A_T$ of different strengths: $A_T$ contains the terminal behaviors, $A_W$ the weak behaviors that are not terminal, and $A_S$ the rest (the strong part).

This is illustrated by Figure 3.1: the initial automaton, at the top, contains four SCCs. $C_1$ is strong, because it mixes accepting and
rejecting cycles. \( C_2 \) is weak, because all its transitions have the same marks. \( C_3 \) is terminal, because all its transitions have the same marks.

\( C_4 \) is rejecting. For any automaton using generalized Büchi acceptance, extracting a subautomaton of a given force can be done by taking all SCCs of that force unchanged, as well as their parents stripped from acceptance marks. For instance, the terminal automaton of Figure 3.1 has a verbatim copy of the terminal component \( C_3 \), plus “naked” versions of its parents \( C_1 \) and \( C_4 \).

This decomposition naturally ensures that \( L(A_{\neg \phi}) = L(A_T) \cup L(A_W) \cup L(A_S) \). In particular, instead of checking \( L(K_M \otimes A_{\neg \phi}) = \emptyset \), a model checker could check \( L(K_M \otimes A_S) = \emptyset \land L(K_M \otimes A_W) = \emptyset \land L(K_M \otimes A_T) = \emptyset \) instead. This has several advantages:

- In the case of \( K_M \otimes A_W \) and \( K_M \otimes A_T \), emptiness checks specialized for weak or terminal automata can be used.
- The three emptiness checks can be parallelized.
- Each one of \( A_S \), \( A_W \), and \( A_T \) is simpler than the original automaton, so it may be easier to simplify than the original (Fig. 3.2). This is particularly true for the weak (and terminal) automata, where the WDBA-minimization discussed on page 13 can be applied.

It should be noted that the languages of these three automata are not necessarily disjoint: some words could be accepted by more than one automaton. Nonetheless each product \( K_M \otimes A_S \), \( K_M \otimes A_W \), and \( K_M \otimes A_T \) is at most as big as the original \( K_M \otimes A_{\neg \phi} \), so testing the emptiness of the former three products by running three model-checkers in parallel can only be faster than checking \( K_M \otimes A_{\neg \phi} \) using one model checker. In our benchmark, we observed speedup factors between 2 and 4 when the products are empty (i.e., the emptiness checks have to construct the entire product). Speedups greater than 3 come from the fact that decomposed automata can be simplified further.

This decomposition technique could be generalized to produce one automaton for each accepting SCC of the original automaton.

3.3 Using Union-find for Emptiness Checks

SCC-based emptiness checks use SCC enumeration algorithms such as the algorithm of Tarjan [1971, 1972] or the lesser known algorithm of Dijkstra [1973, 1976]. For instance, Couvreur [1999] implements Dijkstra’s algorithm (despite the paper presenting it as a variant of Tarjan) while Geldenhuys and Valmari [2004] uses Tarjan.

Another SCC enumeration algorithm was suggested by Gabow [2000, p. 109] and uses the union-find data structure [Tarjan, 1975]. In this algorithm, a DFS of a graph is performed, and the union-find data structure is used to build a partition of the vertices of the graph into SCCs: initially each node is considered to be in its own class of the partition, but whenever a cycle is detected, all states on that cycle are merged into the same class.

In a paper presented at LPAR’13, we described a Union-Find-based emptiness check extending Gabow’s algorithm, and suggested...
This was presented in a sequential context, where the three algorithms give roughly equivalent run times; however our ultimate goal was to investigate these algorithms in a parallel context.

3.4 Parallelisation of SCC-based emptiness checks

Several parallel emptiness checks have been proposed over the last years [Brim et al., 2001, 2004, Černá and Pelánek, 2003, Barnat et al., 2003, 2005, 2009]. These algorithms are mainly BFS-like (or at least "non-DFS") explorations of the state space. Multi-core adaptations of these algorithms with lock-free data structure have been discussed, but not evaluated [Barnat et al., 2010a].

Recent publications show running multiple NDFS-based algorithms with the swarming technique [Holzmann et al., 2011] scale better in practice [Evangelista et al., 2011, Laarman et al., 2011, Laarman and van de Pol, 2011, Evangelista et al., 2012]. In swarming, each thread executes the same algorithm, but uses a different exploration order.

We investigated what could be done with SCC-based emptiness checks in a parallel setting. Although Gabow’s algorithm can be seen as an implementation of Dijkstra’s algorithm using a union-find, we realized that in a parallel context the union-find could serve as a global data structure in which all threads can contribute knowledge about the partition of the states into different SCCs. This shared data structure could be filled by threads running either some variant of Dijkstra or of Tarjan; possibly mixing the two algorithms.

In a parallel setting with \( N \) threads, it makes sense to combine this with the decomposition technique described Section 3.2. We tried that using two different strategies:

**S1:** if the property automaton can be decomposed into \( P \in \{1, 2, 3\} \) automata, then each product will be checked using \( \frac{N}{P} \) threads. In this context the decomposition improves the spread of the swarming technique by ensuring that groups of threads focus on behaviors of different strengths.

**S2:** use all \( N \) threads for the product \( K_M \otimes A_T \), then if no counterexample is found, use \( N \) threads for \( K_M \otimes A_W \), and finally \( K_M \otimes A_S \). In this strategy, the automata are ordered by strengths so that more complex emptiness checks are avoided when possible.

Figure 3.3 shows the effect of these two strategies compared to our base parallelization of Dijkstra, in a benchmark derived from BEEM [Pelánek, 2007]. Strategy S1 is not used with 1 or 2 threads, as the decomposition could produce 3 automata. Nonetheless, S1 appears good in both empty products (the full product had to be explored to prove the formula satisfied) and non-empty product (one thread found a counterexample and other threads could be stopped immediately).
4

Generic acceptance

So far, we have discussed automata with generalized Büchi acceptance, or subclasses of this acceptance (like plain Büchi). In this chapter we enlarge the notion of acceptance condition to also include acceptance conditions such as co-Büchi, Rabin, Streett, or any Boolean combination of those.

4.1 The HOA format

We shall specify the acceptance condition of an automaton as a formula, using a syntax introduced in the Hanoi Omega Automaton (HOA) format\(^1\). In this format, the acceptance condition is a formula over terms such as \(\text{Inf}(x)\) or \(\text{Fin}(x)\) indicating respectively that for a run to be accepting the acceptance mark \(x\) should be visited infinitely often, or finitely often.\(^2\)

Table 4.1 shows some examples.

<table>
<thead>
<tr>
<th>Büchi:</th>
<th>(\text{Inf}(0))</th>
</tr>
</thead>
<tbody>
<tr>
<td>generalized Büchi 3:</td>
<td>(\text{Inf}(0) \land \text{Inf}(1) \land \text{Inf}(2))</td>
</tr>
<tr>
<td>co-Büchi:</td>
<td>(\text{Fin}(0))</td>
</tr>
<tr>
<td>generalized co-Büchi 3:</td>
<td>(\text{Fin}(0) \lor \text{Fin}(1) \lor \text{Fin}(2))</td>
</tr>
<tr>
<td>Rabin (2 pairs):</td>
<td>((\text{Fin}(0) \land \text{Inf}(1)) \lor (\text{Fin}(2) \land \text{Inf}(3)))</td>
</tr>
<tr>
<td>Streett (2 pairs):</td>
<td>((\text{Fin}(0) \land \text{Inf}(1)) \land (\text{Fin}(2) \lor \text{Fin}(3)))</td>
</tr>
<tr>
<td>parity min even 5:</td>
<td>(\text{Inf}(0) \lor (\text{Fin}(1) \land (\text{Inf}(2) \lor (\text{Fin}(3) \land \text{Inf}(4)))))</td>
</tr>
</tbody>
</table>

The HOA format was created jointly with other tools authors, as a way to facilitate the interactions between our tools. Existing formats, such as the DSTAR format of \(\text{ltl2dstar}\) \cite{Klein:Baier:2006} or the XML-based format of \(\text{Goal}\) \cite{Tsai:2013} support different acceptance conditions that can only be selected from a predefined list of names. One particular motivation for HOA was that tools like \(\text{ltl3dra}\) \cite{Babiak:2013} and \(\text{Rabinizer}\) \cite{Kretinsky:Esparza:2012} had introduced a new acceptance condition that they called \textit{generalized Rabin} and were each using a custom output format for this new type of automata. The HOA format was designed not to be restricted to a list of known acceptance conditions: specifying the acceptance condition using a formula rather than by name, so that the semantics are always known.

---


\(^2\) A similar formalism was already used long ago by Emerson and Lei \cite{Emerson:Lei:1987}: they write \(\mathcal{F}(p_0) \land \mathcal{G}(p_1)\) when we write \(\text{Inf}(0) \land \text{Fin}(1)\). Safra and Vardi \cite{Safra:Vardi:1989} named automata with similar acceptance conditions “EL automata”.

---

Table 4.1: Example of traditional acceptance conditions, specified using the HOA syntax.

```
$ ltltgba 'a U b'
HOA: v1
name: "a U b"
States: 2
Start: 1
AP: 2 "a" "b"
acc-name: Buchi
Acceptance: 1 \text{Inf}(0)
properties: trans-labels
        explicit-labels state-acc
deterministic
stutter-invariant terminal
--BODY--
State: 0 (0)
[t] 0
State: 1
[1] 0
[0&!1] 1
--END--
```

Example 4.2: A Büchi automaton generated for \(a U b\) in the HOA format, and its graphical representation.
4.2 Why use generic acceptance?

First, it should be pointed out that using more complex acceptance conditions does not necessarily imply slower procedures. In the context of model checking, we have already discussed the fact that using generalized-\Buchi instead of \Buchi could be faster.\(^3\) Similarly, it has been shown that using the generalized Rabin condition can speed up probabilistic model checking by orders of magnitude [Chatterjee et al., 2013, Komárková and Křetínský, 2014].

Second, not being tied to a particular acceptance condition makes some \omega-automata operations easier. For instance, the product or union of two deterministic (and complete) automata is trivial to define without the constraint of producing the same acceptance (as we will discuss below).

Finally, providing algorithms that support generic acceptance opens up a wide research area, the results of which can also benefit to people working on specific subclasses of acceptance.

4.3 Transition-based acceptance revisited

In section 2.4 we discussed the fact that transition-based acceptance was not useful to weak \Buchi automata, and therefore to all persistence properties. That is true as long as it is clear we are talking about automata with \Buchi acceptance (even generalized). However, persistence properties can benefit from transition-based acceptance if we consider acceptance conditions involving Fin(x). The typical example is the property FG that can be reduced to a 1-state (deterministic) transition-based co-\Buchi automaton (Fig. 4.3).

4.4 Generic algorithms

Figure 4.4 shows two deterministic and complete Streett automata A and B, along with their synchronous product. If we name C_1 the product equipped with the acceptance condition (Fin(x) \lor Inf(x)) \land (Fin(x) \land Inf(x)), then C_1 is a Streett automaton verifying \(L(C_1) = L(A) \cap L(B)\). However, if we call C_2 the product equipped with the acceptance condition Fin(x) \lor Inf(x) \lor Inf(x), then C_2 verifies \(L(C_2) = L(A) \lor L(B)\). Note that the acceptance condition of C_2 is not Streett, and does not correspond to any “traditional” acceptance condition.

This construction for the intersection or union of two \omega-automata works regardless of the acceptance condition, and also on non-deterministic automata. The union only requires complete automata.

Complementing a deterministic and complete automaton is as simple as complementing its acceptance condition and applying the rules \(\neg \text{Fin}(x) \equiv \text{Inf}(x)\) and \(\neg \text{Inf}(x) \equiv \text{Fin}(x)\).

Completing an \omega-automaton is trickier than it sounds: after a sink state is added, it should be given a set of acceptance marks that does not satisfy the acceptance condition. An automaton can...
therefore be completed without changing its acceptance condition unless that acceptance is a tautology. If the acceptance is a tautology, like $\text{Fin}(0) \lor \text{Inf}(0)$, then the input automaton contains only accepting cycles, and the output automaton can be given the Büchi condition in order to mark all states except the sink as accepting.

Many of the simplification algorithms that we discussed in Chapter 2 (such as simulation-based reductions, WDBA minimization, SCC-based simplifications) can be easily adapted to generic acceptance: in Spot they are therefore implemented for automata with arbitrary acceptance (Example 4.6).

For algorithms that are harder to generalize (e.g., determinization, emptiness checks), it is always possible to build an equivalent automaton with a more constrained acceptance condition, as discussed in the next section.

### 4.5 Fin-less acceptance

The SCC-based emptiness-check discussed in Chapter 3 were presented in the context of automata using generalized Büchi acceptance. However, these emptiness checks will also work with automata whose acceptance condition is a disjunction of generalized Büchi acceptance conditions. We call this subclass of acceptance conditions “Fin-less acceptance”, because it is simply a positive Boolean combination of $\text{Inf}(x)$ terms.

A simple transformation can be used to transform an automaton with arbitrary acceptance condition into an automaton with Fin-less acceptance. Let us assume, without loss of generality\(^4\), that the input acceptance condition is under the following disjunctive normal form:

$$
\bigvee_i (\text{Fin}(x_{i1}) \land \ldots \land \text{Fin}(x_{in_i}) \land \text{Inf}(y_{i1}) \land \ldots \land \text{Inf}(y_{in_i}))
$$

where $n_i \geq 0$ and $m_i \geq 1$ for all $i$, and that the sets of $\text{Inf}$ marks of each clause $Y_i = \{y_{i1}, \ldots, y_{in_i}\}$ are such that no set is a superset of another: $\forall i, j, Y_i \subseteq Y_j$.

Under these assumptions, the input automaton can be transformed into another automaton with $\bigvee_i (\text{Inf}(y_{i1}) \land \ldots \land \text{Inf}(y_{in_i}))$ as acceptance condition. The transformation is as follows: For each SCC $C_j$ of the original automaton and each clause $i$ whose $\text{Fin}$ marks are present in $C_j$, create a clone $D_{ij}$ of the SCC $C_j$, remove from that clone all transitions (or states) that have some marks in $X_i = \{x_{i1}, \ldots, x_{in_i}\}$, and add at least one non-deterministic jump from $C_j$ to $D_{ij}$ per elementary cycle of $D_{ij}$. Finally, in the SCCs that correspond to the original automaton, leave only the marks that belong to some $X_i$ for which $Y_i = \emptyset$.

To avoid the complexity of enumerating the elementary cycles of $D_{ij}$, our implementation creates one non-deterministic jump each time a back-edge is found during a depth-first search of $C_j$. This potentially creates more non-deterministic jumps than needed. Fig. 4.7 gives an example of such transformation (note that the bottom right state generated by this construction is superfluous).

---

\(\text{Fin}(0) \lor \text{Inf}(0)\) translates $G_a$ into a 3-state complete deterministic state-based Rabin automaton. Passing this automaton to autfilt for simplification with instructions to output a deterministic (-D) and complete (-C) automaton with state-based acceptance (-S), yields a 2-state automaton. In this case, autfilt applied the algorithm of Dax et al. [2007] (generalized to work on Rabin automata) to detect that the input automaton represents an obligation property and can therefore be minimized into a 1-state weak deterministic Büchi automaton. This automaton has 2 states once completed.

---

\(^4\) If the input acceptance does not satisfy these assumptions, it can be easily modified by adding some $\text{Inf}(z)$ terms and adding the mark $z$ everywhere.

---

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

---

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

---

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

---

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

---

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$

---

\$\text{echo G a | ltl2dstar -- |}\
> \text{autfilt --remove-fin in >out}\$
Spot implements a variant of this construction that guarantees that if the input acceptance needs to be “fixed” to satisfy the assumptions, the output acceptance uses at most one additional mark. Additionally, we have implemented some transformations that are specific to some subclasses of acceptance conditions and that are automatically used when applicable: for instance state-based Rabin automata can be converted to Büchi automata, preserving determinism whenever possible, using a variant of an algorithm by Krishnan et al. [1994].

As already mentioned, automata with Fin-less acceptance conditions can be easily checked for emptiness using readily available emptiness checks, such as those discussed in Chapter 3. Those emptiness checks are not appropriate for automata that use Fin. The emptiness check of automata with generic acceptance has been considered long ago: Emerson and Lei [1985] studied the problem of detecting fair states in a finite system under a fairness hypotheses defined as a Boolean function using operators similar to Inf and Fin. Their result translates immediately to the emptiness check problem for \( \omega \)-automata: emptiness check with generalized acceptance is NP-complete, but they provide a polynomial algorithm for a large subclass of acceptance conditions: all conditions that can be encoded as a disjunction of Strerett acceptance (this encompasses Rabin and generalized-Rabin).

We plan to investigate these “generic emptiness checks” in the future. In the meantime, using our Fin-removal procedure in front of more traditional emptiness checks is enough to provide a usable system. The only part of our procedure that does not have a polynomial complexity is the conversion of the acceptance condition into conjunctive normal form, which we implement using binary decision diagrams to prune redundant terms.

Finally, it should be noted that a Fin-less acceptance can be converted into generalized Büchi acceptance by first converting it into conjunctive normal form, and then replacing each clause \( \text{Inf}(x_{1i}) \vee \ldots \vee \text{Inf}(x_{ni}) \) by a single \( \text{Inf}(y_i) \) where \( y_i \) is a mark that should label all transitions (or states) marked with any of \( x_{1i}, \ldots, x_{ni} \). In the case of the automaton of Figure 4.7, the resulting acceptance is already a single-clause CNF, so it can be trivially converted into Büchi acceptance as in Figure 4.8.

\[ \text{autfilt --tgba in >out} \]

\[ \text{Figure 4.8: Converting the automaton of Figure 4.7 to use generalized Büchi acceptance, or in this case, Büchi acceptance.} \]

4.6 Example tool benefiting from generic acceptance conditions

Spot distributes a tool called ltlcross\(^5\), that started as a reimplementaton of the no-longer-maintained lbtt: a randomized test-bench for translators of LTL formulas into (generalized) Büchi automata [Tauriainen and Heljanko, 2002]. The current version of ltlcross of course supports arbitrary acceptance conditions. It is regularly used to test Spot’s translation algorithms, but also has been used for instance by the authors of ltl3ba [Babiak et al., 2012], ltl3dra [Babiak et al., 2013b], ltl2dstar [Klein and Baier, 2006, 2007], or Rabinizer3 [Komárová and Křetínský, 2014] to test M. Duret-Lutz. Manipulating LTL formulas using Spot 1.0. In ATVA’13, vol. 8172 of LNCS, pp. 442–445. Springer, 2013

\[ \text{$autfilt --tgba in >out$} \]

\[ \text{Figure 4.8: Converting the automaton of Figure 4.7 to use generalized Büchi acceptance, or in this case, Büchi acceptance.} \]

\[ \text{Figure 4.8: Converting the automaton of Figure 4.7 to use generalized Büchi acceptance, or in this case, Büchi acceptance.} \]
Recent versions of their tools.

Given a formula $\phi$ and a list of translators $T_1, \ldots, T_m$ producing $\omega$-automata from formulas, ltlcross calls each tool on $\phi$ and its negation, yielding a positive automaton $P_i = T_i(\phi)$ and a negative automaton $N_i = T_i(\neg \phi)$.

Now ltlcross (like lbtt did for generalized Büchi) will make sure that $\mathcal{L}(P_i \otimes N_j) = \emptyset$ for all $i$ and $j$. It could be the case that $P_i$ is a Rabin automaton, and that $N_j$ is a Streett automaton, but clearly building the product $P_i \otimes N_j$ and testing its emptiness for arbitrary acceptance condition can be done using the techniques discussed in this chapter. However, this test is not enough to prove that all translators produce equivalent automata: for instance if one translator always returns empty automata, this test will pass. Therefore ltlcross also tests $\mathcal{L}(P_i \otimes N_i) = \emptyset$: when $P_i$ and $N_i$ are deterministic, complementing them is as simple as completing the automaton and complementing its acceptance condition. For non-deterministic automata, ltlcross offers the (potentially costly) option to determinize them, as illustrated by Figure 4.9.

Our determination procedure currently takes transition-based Büchi automata as input (so we may have some preprocessing to do if the input has a different acceptance), and outputs automata with transition-based parity acceptance. It mixes the construction of Redziejowski [2012] with some optimizations of ltl2dstar [Klein and Baier, 2006, 2007] and a few of our own. It would be nice to be able to determine automata with arbitrary acceptance conditions; Varghese [2014] offers some clues.

Yet unpublished.
5
SAT-based minimization of deterministic automata

LTL Synthesis [Finkbeiner and Schewe, 2005] and Probabilistic LTL
Model Checking [Baier and Katoen, 2008] are two areas where it is
useful to express linear-time temporal properties as deterministic
ω-automata. Because it is well known that not all Büchi automata
can be made deterministic, these applications use other acceptance
conditions such as Rabin or Streett.

In this chapter we discuss the minimization of such deterministic
automata, with arbitrary acceptance conditions. This is the result of
an ongoing collaboration with Soheib Baarir.

5.1 Existing results

Minimizing deterministic Büchi automata, deterministic co-Büchi
automata, and deterministic parity automata is known to be NP-
complete [Schewe, 2010], so we should not hope for efficient mini-
mizations that work for arbitrarily complex acceptance conditions.

However, and as already pointed out on page 13, minimizing
automata that are weak and deterministic can be done in polynomial
time [Löding, 2001]. For this subclass of automata, it does not even
matter what the acceptance condition is: any (inherently) weak and
deterministic can be rewritten as a Büchi or co-Büchi automaton
without changing the transition structure.

For more complex automata, we do not know of efficient min-
imization procedures. Tools usually apply simplifications that do
not guarantee a minimal result. Being able to see how far those
simplifications are from the minimal automata would provide a good
help to improve them.

5.2 Minimization via SAT-solving

A technique (and tool) for minimizing (state-based) deterministic
Büchi automata was first presented by Ehlers [2010]. His tool, called
DBAminimizer, will take deterministic Rabin automata produced by
ltl2dstar, will convert it into a deterministic Büchi automaton if it
exists [Krishnan et al., 1994], and finally will minimize the resulting
Büchi automaton using a SAT-solver.

The tool implements a minimization loop, in which each iteration
tries to synthesize an automaton that has one less state than the previous automaton.

The SAT-based synthesis of an equivalent automaton works as follows. A set of variables denoting the existence of each possible transition and the acceptance of each state in the “candidate” automaton. These variables are used to derive the values of another set of variables encoding a product between the reference automaton, and the candidate automaton. Finally, constraints are added such that for each cycle of the product automaton, its projection on the reference automaton is accepting if and only if its projection on the candidate automaton is also accepting.

5.3 Our improvements

We added several improvements to the above minimization technique when we implemented it in Spot.

• First, Spot is able to detect obligation properties, so we can avoid SAT-solvers and use a polynomial minimization in this case.

• Second, while Spot will not always produce a deterministic Büchi automaton from an LTL formula when such automaton exists, it will often do so. And since the deterministic automata produced this way are usually much smaller than the Rabin automata produced by ltl2dstar from determinization, it made sense to first try that, and to adapt the minimization technique to transition-based acceptance.

• Third, we improve Ehlers [2010]’s encoding by adding several optimizations to reduce the number of variables and clauses needed by the encoding. Some of these optimizations are based on the knowledge of the strongly connected components of the original automaton (there is no need to track cycles outside of those), and knowing which strongly connected components of the original automaton are weak.

• Finally, we generalize the technique to deal with transition-based generalized Büchi acceptance\(^1\), and then arbitrary acceptance conditions\(^2\). These generalizations come at the cost of additional clauses and variables.

Currently, the implementation has two different encodings: a simple one that works with deterministic transition-based Büchi automata as input and output, and a more general encoding that can input deterministic automata with an arbitrary acceptance, and output possibly another arbitrary acceptance. Both encodings can be configured to use state-based acceptance if needed.

Figure 5.1 shows how a 4-state deterministic Rabin automaton constructed by ltl2dstar can be minimized into a deterministic Rabin automaton with 3 states if state-based acceptance is used, or just 1-state if transition-based acceptance is used. The bottom

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example of Figure 5.1 shows how we can also force an arbitrary acceptance condition on the output: here we want parity acceptance with numbers 0, 1, 2 such that a run is accepting if the maximum number seen infinitely often is even. The colored option constrains each state to belong to exactly one acceptance set (this is translated into additional constraints for the SAT solver), which is what one expects from a parity automaton.

5.4 Applications

Let us assume an alphabet of size $|\Sigma|$. To minimize an automaton that uses an $n$ states and $m$ acceptance sets into another automaton that uses $n'$ states and $m'$ acceptance sets, our SAT encoding requires at most $O(n^2 \times n'^2 \times 2^{m+m'})$ variables and $O(n^2 \times n'^3 \times 2^{2m+m'} \times |\Sigma|)$ clauses.

For this reason the minimization procedure is not very practical for automata with a large number of states, or more that 5–6 acceptance sets. However, automata used in model checking are usually quite small, so attempting a SAT-based minimization can be worth a try.

Our main use case however, is to “mine” sub-optimal automata in the output of other tools. Indeed, it is easier to improve an LTL translator when you are provided with a list of formulas for which the output is far from optimal. Our papers contain experiments showing lists of formulas for which existing translators fail to produce reasonably-sized automata: looking at those automata can suggest improvements to the algorithms.

As a simple example, we discovered that Spot translates the LTL formula $\phi = (GF(a \land GF(b)) \lor GF(c)$ into a 3-state non-deterministic transition-based generalized Büchi automaton with 2 acceptance sets (Fig. 5.2). However, after determinizing this automaton and minimizing it, we discovered that there exists a 1-state deterministic automaton with the same acceptance. The equivalent formula $GF(a \lor c) \land GF(b \lor c)$ is translated to the minimal automaton directly.

This suggests that implementing LTL simplification rules that distribute through GF terms would be an improvement. But that is not the only option.

Since this technique works with arbitrary acceptance conditions as input and output, it can also be used to search of sub-optimal automata produced by other algorithms such as acceptance conversions, and other transformations.
An $\omega$-regular language is stutter-invariant if it is closed under the operation that duplicates some letter in a word or that removes some duplicate letter [Etessami, 1999].

We say that a formula is stutter-invariant if its language is. Stutter-invariant formula form an important subclass of temporal properties, as they enable model checkers to apply partial-order reduction techniques (e.g., [Clarke et al., 2000, Ch. 10] or [Baier and Katoen, 2008, Ch. 8]) to reduce the set of behaviors they must explore. Such partial-order reductions are implemented by explicit model checkers such as Spin [Holzmann, 2003, Ch. 9], LTSMIN [Laarman et al., 2014], or DiVinE [Barnat et al., 2010b], to cite a few. Detecting stutter-invariant properties has also uses beyond partial-order reductions; for instance it is used to optimize the determinization constructed in ltl2dstar [Klein and Baier, 2007].

In this chapter, we present two results related to stutter-invariance: the first is an efficient algorithm for deciding whether a property is stutter-invariant, the second is a generalization of a type of automata known as “testing automata” that can only represent stutter-invariant properties.

6.1 Stutter-invariance checks

It is widely known that any LTL formula that does not use the next-step operator $X$ (a.k.a. an LTL $\setminus X$ formula) is stutter-invariant; this check is trivial to implement. Unfortunately there exist formulas using X that are stutter-invariant and whose usage is desirable [Păun and Chechik, 2003]. For instance, the stutter-invariant formula $F(a \land X(\neg a \land b))$ specifies that $b$ will be true at a moment where $a$ was just switched off.

Dallien and MacCaull [2006] built a tool that recognizes a stuttering LTL formula if (and only if) it matches one of the patterns of Păun and Chechik [2003]. This syntactical approach is efficient, but incomplete, as not all stutter-invariant formulas follow the recognized patterns.

A more definite procedure was given by Peled and Wilke [1997] as a construction that inputs an LTL formula $\varphi$ with $|\varphi|$ symbols and $n$ atomic propositions, and outputs an LTL $\setminus X$ formula $\varphi'$ with $O(4^n |\varphi|)$ symbols.

\begin{verbatim}
6 ltl2dstar & remove-x \\
> -f 'lts2min(1a2b)'
- (a \land (b \lor \neg a) \land (b \lor \neg a \land b)) \\
> -f 'lts2min(1a2b)'
- (a \land (b \lor \neg a) \land (b \lor \neg a \land b)) \\
\end{verbatim}
symbols, such that \( \varphi \) and \( \varphi' \) are equivalent iff they represent a stutter-invariant property. This construction, which proves that any stutter-invariant formula can be expressed without \( X \), was later improved to \( n^{O(k)}|\varphi| \) symbols, where \( k \) is the \( X \)-depth of \( \varphi \), by Etessami [2000] (see Example 6.1). If a disjunctive normal form is desired, Tian and Duan [2009] give a variant with size \( O(n^2|\varphi|) \). To decide if an LTL formula \( \varphi \) is stutter-invariant, we build \( \varphi' \) using one of these constructions, and then check the equivalence of \( \varphi \) and \( \varphi' \). This equivalence check can be achieved by translating these formulas into automata.

This approach, based on Etessami’s procedure, was implemented in old versions of Spot\(^2\) (and can still be performed explicitly, as in Example 6.2). However, two issues motivated us to look into alternative directions:

- As seen in Example 6.1, Etessami’s rewriting function (called \( \tau^2 \)) can generate very large formulas. Testing equivalence via translation of these formulas can therefore be very costly.

- Etessami’s procedure only works for LTL, but we wanted to decide the stutter invariance of PSL\(^2\) properties as well. Dax et al. [2009], proposed a rewriting technique that is to PSL what Etessami’s procedure is to LTL, however after implementing it we discovered it was incorrect, and no fix has been found so far.

Instead we took inspiration from the procedure used in \texttt{ltl2dstar} [Klein and Baier, 2007], where stutter-invariance is decided at the automaton level. Given an formula \( \varphi \), they first build an equivalent Büchi automaton \( A_\varphi \), then they construct an automaton \( A'_\varphi \) that accepts the smallest stutter-invariant language over-approximating the language of \( \varphi \). The property \( \varphi \) is stutter-invariant iff \( A_\varphi \) and \( A'_\varphi \) have the same language, which can be checked by ensuring that the product \( A'_\varphi \otimes A_\varphi \) has an empty language.

In a paper published at \textsc{Spin’15}\(^3\), we proposed several alternatives to this construction.

An easy intuition can be provided by seeing the construction of \( A'_\varphi \) as \( A'_\varphi \equiv \text{cl}(\text{sl}(A_\varphi)) \), where \( \text{sl} \) (for “self-loopize”) transforms the automaton so it accepts all the original words plus those you can obtain by duplicating some letters, and \( \text{cl} \) (for “closure”) transforms the automaton so it additionally accepts words obtained by omitting duplicate letters (Fig 6.3).

Because \( \mathcal{L}(\text{cl}(\text{sl}(A_\varphi))) \subseteq \mathcal{L}(A_\varphi) \) by construction, testing the equivalence of the two automata can be done by testing just that \( \mathcal{L}(A_\varphi \otimes \text{cl}(\text{sl}(A_\varphi))) = \emptyset \). In this setup, our main result can be expressed as follows:

\textbf{Theorem S.} Let \( \varphi \) be a property expressed as a \( \omega \)-automaton \( A \), and assume we know how to obtain \( \overline{A} \). Testing \( \varphi \) for stutter-invariance is equivalent to testing the emptiness of any of the following products: \( \text{cl}(\text{sl}(A)) \otimes \overline{A} \), \( \text{sl}(\text{sl}(A)) \otimes \overline{A} \), \( \text{sl}_2(\text{cl}(A)) \otimes \overline{A} \), \( \text{sl}_2(\text{sl}_2(\text{cl}(A))) \otimes \overline{A} \), \( \text{sl}_2(\overline{A}) \otimes \text{sl}_2(\overline{A}) \), or \( \text{cl}(A) \otimes \text{cl}(\overline{A}) \).

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2 Cf. Sec. 7.2 p. 40.


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Figure 6.3: Examples of application of the \texttt{cl} and \texttt{sl} operators.
which I co-supervised with F. Kordon. However, the presentation of their time translating the LTL formulas in automata, so it is better to keep these formulas short.

If we ignore the translation time, which is common to all the checks mentioned in Theorem S, we observe that $\mathcal{L}(\text{cl}(A_\varphi) \otimes \text{cl}(A_{\neg \varphi})) = \emptyset$ is the most efficient in our implementation (see Table 6.6). As a consequence, this is now the default stutter-invariance check in Spot.

### 6.2 Generalized Testing Automata

This section was the subject of A. E. Ben Salem’s Ph.D. thesis [2014] which I co-supervised with F. Kordon. However, the presentation of the material, using the notations of Chapter 4 to discuss the

| $|AP| = 1$ | $|AP| = 2$ | $|AP| = 3$ |
|-----------|-----------|-----------|
| $\mathcal{L}(A_{\tau(\varphi)} \otimes A_{\neg \varphi}) = \emptyset \land \mathcal{L}(A_{\neg \tau(\varphi)} \otimes A_\varphi) = \emptyset$ | $0.32s$ | $40.62s$ | $>4801s$ (OOM) |
| $\mathcal{L}(A_{\neg \tau(\varphi)}) = \emptyset$ | $1.18s$ | $3347.92s$ |
| $\mathcal{L}(\text{cl}(\text{sl}(A_\varphi)) \otimes A_{\neg \varphi}) = \emptyset$ | $0.61s$ | $1.91s$ | $6.14s$ |
| $\mathcal{L}(\text{sl}(\text{cl}(A_\varphi)) \otimes A_{\neg \varphi}) = \emptyset$ | $0.61s$ | $1.91s$ | $6.10s$ |
| $\mathcal{L}(\text{cl}(\text{sl}_2(A_\varphi)) \otimes A_{\neg \varphi}) = \emptyset$ | $0.61s$ | $1.89s$ | $5.97s$ |
| $\mathcal{L}(\text{sl}_2(\text{cl}(A_\varphi)) \otimes A_{\neg \varphi}) = \emptyset$ | $0.61s$ | $1.91s$ | $5.97s$ |
| $\mathcal{L}(\text{sl}(A_\varphi) \otimes \text{sl}(A_{\neg \varphi})) = \emptyset$ | $0.61s$ | $1.92s$ | $6.18s$ |
| $\mathcal{L}(\text{sl}_2(A_\varphi) \otimes \text{sl}_2(A_{\neg \varphi})) = \emptyset$ | $0.61s$ | $1.90s$ | $5.99s$ |
| $\mathcal{L}(\text{cl}(A_\varphi) \otimes \text{cl}(A_{\neg \varphi})) = \emptyset$ | $0.60s$ | $1.89s$ | $5.94s$ |

number of stutter-invariant formulas found | 234 | 162 | 112 |

Table 6.4: Time to classify 500 random LTL formulas that all use the X operator and have the given number of atomic propositions.

Table 6.6: Cross-comparison of the checks of Theorem S on 40000 random LTL formulas with X. A value $v$ on line $x$ and column $y$ indicates that there are $v$ cases where check $(x)$ was more than 10% slower than check $(y)$. In other words, a line with many small numbers indicates a check that is usually faster than the others.

In this theorem, sl$_2$ is an alternative construction that produces automata with the same language as those output by sl.

The surprise of this theorem is in the last three products: it is indeed possible to test stutter-invariance by implementing any one of the three functions cl, sl, sl$_2$, while previous work is equivalent to a combination of two [Klein and Baier, 2007].

In the context of deciding stutter invariance for LTL formulas, our experimental benchmark in the case of LTL formulas showed all the decision procedures of this theorem are orders-of-magnitude faster than our previous decision procedure based on Etessami’s rewriting (Table 6.4). The reason is that all these procedures spend most of their time translating the LTL formulas in automata, so it is better to keep these formulas short.

If we ignore the translation time, which is common to all the checks mentioned in Theorem S, we observe that $\mathcal{L}(\text{cl}(A_\varphi) \otimes \text{cl}(A_{\neg \varphi})) = \emptyset$ is the most efficient in our implementation (see Table 6.6). As a consequence, this is now the default stutter-invariance check in Spot.

Example 6.5: Continuing the example of Fig 6.1 and Fig 6.2: the option filters stutter-invariant formulas using the last check of Theorem 5.

**Table 6.4:**

\[
\text{lctlfilt --stutter-inv} \\
\text{> -f "$E^a -E \neg \neg E^g"} \\
F(a \& X(!a \& b))
\]

**Example 6.5:**

Cross-comparison of the checks of Theorem 5 on 40000 random LTL formulas with X. A value $v$ on line $(x)$ and column $(y)$ indicates that there are $v$ cases where check $(x)$ was more than 10% slower than check $(y)$. In other words, a line with many small numbers indicates a check that is usually faster than the others.
acceptance condition, is new.

Testing automata were introduced by Hansen et al. [2002] as an alternative to Büchi automata for the automata theoretic approach to model checking (Fig. 1.1, p. 5). Contrary to Büchi automata which test the values of the different atomic propositions of the system (either “light is on” or “light is off”), testing automata may only detect the changes of those values (“the status of light was changed”). Testing automata should remain in their current state as long as none of the observed properties of the system change. This has two consequences: first, testing-automata naturally capture only stutter-invariant properties, and second, they need two forms of accepting runs. They use a Büchi acceptance condition to accept some runs where atomic propositions are changing infinitely often, and another “livelock acceptance condition” to capture runs where no observed atomic proposition change.

Figure 6.7 shows an example of testing automata and presents the two different acceptances. Because the two forms of acceptance have to distinguish runs that stutter continuously from those that do not, the acceptance condition cannot precisely be expressed in terms of the Fin and Inf operators introduced in Chapter 4. However, if we make the stuttering explicit in the automaton and use transition-based acceptance, as in Figure 6.8, then the acceptance condition is similar to a Streett condition with a single pair.

Algorithms for constructing testing automata from LTL formulas (via Büchi automata), for constructing the product between a Kripke structure and a testing automaton, and for testing their emptiness were discussed by Hansen et al. [2002] and Geldenhuys and Hansen [2006]. In the general case, the emptiness check requires two passes: one pass for testing Büchi acceptance, and one pass for testing livelock acceptance. However, to prove the absence of ∅-labeled cycles, one has to make a dedicated pass on the product between the Kripke structure and the testing automaton.

Our first contribution was to implement testing automata, and reproduce the results observed by Geldenhuys and Hansen [2006]: although testing automata are usually larger than the corresponding Büchi automata, they usually yield smaller products with the system, and counterexamples are found quicker on average.4 However, proving the absence of counterexample takes more time due to the 2-pass emptiness check. We also compared the testing automaton approach against one using transition-based generalized Büchi automata and found similar results.

However, once the stuttering self-loops are explicitly represented, one can more easily realize that the Fin(¬∅) acceptance can be removed from the automaton by adding a new state dedicated to accepting stuttering behaviors, and making non-deterministic jumps into this state (Fig. 6.9).

This construction should look similar to the Fin-removal of Section 4.5, except we introduce non-deterministic jumps before entering a livelock state so that the non-deterministic jumps are not

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labeled by ∅: the only reason for that is to stay compatible with the testing automaton formalism, where stuttering should not move the automaton to another state. Note that states that are both Büchi accepting and livelock accepting can be fixed without introducing a new transition.

In practice, the product between these 1-pass testing automata and a Kripke structure can be seen as a Büchi automaton labeled by changesets, and can be checked for emptiness using any Büchi emptiness check, because emptiness checks do not look at labels. The second pass is no longer needed, since livelock acceptance is not used anymore. However, the extra state and non-deterministic transitions introduced in the automaton will synchronize with the same part of the Kripke structure that a second pass would have to explore; however since everything is explored at once, it requires more memory. Experiments show that 1-pass testing automata are comparable to classical testing automata as far as run time goes [Ben Salem, 2014, Sec. 4.4]. This is therefore a trade-off between ease of implementation (reusing existing emptiness checks for Büchi acceptance) and memory footprint.

We generalized 1-pass testing automata in a very natural way: instead of constructing them from state-based Büchi automata, we generate them from transition-based generalized Büchi automata, keeping the generalized Büchi acceptance. The result, which we call Transition-based Generalized Testing Automata (TGTA), combines the advantages of TGBAs (the conciseness illustrated on Fig. 2.1(d), p. 9) with the use of changeset labels from testing automata. Furthermore, because we create TGTA from TGBA, which are transition-based, we can output smaller automata even when when generalized Büchi is not used. Figure 6.10 shows an example.

Experiments reveal that TGTA are generally better than TGBA, as seen on Figure 6.11. On empty products (i.e., verified properties) emptiness checks using TGTA visit on the average 17.3% fewer transitions than emptiness checks using TGBAs, and 13.4% fewer transitions than those using TAs. For non-empty products (i.e., a counterexample exists), the picture is less clear, due to the fact that a different order of transitions can help to find a counterexample.

\[ \text{Figure 6.10: A Transition-based Generalized Testing Automaton for } a \lor Gb. \text{ Here the acceptance condition use a single Büchi set, but it could use more.} \]

\[ \text{Figure 6.11: Comparison of the TGTA approach against the TA and TGBA approaches on a sample of 2049 pairs (model, formula) where the formula is verified (•) and 2049 pairs (model, formula) where the formula is violated (●). We compare the number of transitions visited by the emptiness checks. TGTA are better for all the dots below the diagonal. Refer to Ben Salem et al. [2012] for the experimental setup.} \]
earlier: the differences can be so important that averaging them would make no sense. Let us just notice than the TGTA approach looses over the TGBA approach in nearly 39% of the non-empty (●) cases.

The above experiment, done in the context of explicit model checking was also successfully transposed to a symbolic setting and produced similar results. In symbolic model checking, all variants of testing automata offer the nice guarantee that one can always stutter on any state. This can be used by saturation techniques [Ciardo et al., 2003, Thierry-Mieg et al., 2009] to speed up the construction of the stuttering part of the product.

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7

Conclusion

7.1 Summary and Perspectives

In this document, I have presented a selected subset of the work I achieved since 2007. Some topics have been left out for reasons that will be discussed in Section 7.2.

Most of the contributions can probably be described as “practical improvements to a well-known theory”. This comes from the fact that my motivation is to implement a toolbox (Spot) of efficient and reusable algorithms that are useful to researchers and teachers in the domain of $\omega$-automata and model checking.

Let us quickly summarize and present perspectives for each chapter.

Chapter 2 presented all the techniques that have been implemented to make ltl2tgba one of the best tools to translate LTL formulas into small automata with few nondeterministic choices. While we do not believe there is a lot of room for improvement in the core algorithm used for translation, we think that the pre- and post-processings can be improved. Firstly, section 5.4 already hinted that better LTL simplifications could be found. Secondly, the simulation-based automata-reduction algorithm currently implemented only use “direct” or “backward” simulation-relations, but there exist coarser relations that can also be used for quotienting [Mayr and Clemente, 2013]. Most of these coarser relations have only been studied in the context of Büchi automata, or maybe generalized Büchi automata, but it would be interesting to extend those to generalized acceptance. Lastly, we should keep in mind that most LTL translators so far have tried to produce smaller automata or more deterministic automata, but we have shown (Section 2.9) that it makes sense to tune the generation of these automata according to how they will be used (e.g., the location of accepting states can affect model checkers using emptiness checks based on a nested-DFS). Working on translations from LTL to automata with different acceptance conditions (not generalized Büchi) is another interesting perspective, especially to produce deterministic automata, which is desirable for probabilistic model checking.

Chapter 3 presented some new ideas to parallelize emptiness checks for generalized Büchi automata. One is to decompose the
property automaton in sub-automata of different strengths that can each be checked using an emptiness check algorithm tailored to this particular strength. This decomposition could in fact be pushed further in automata with many SCCs: there is no reason to restrict ourselves to one automaton per strength. A second idea was to use the union-find data structure to implement an SCC-based emptiness check: such a data structure is well-suited to be shared between multiple threads that explore the automaton. This work was the subject of Étienne Renault’s Ph.D. thesis.

Chapter 4 discussed our adoption of automata with generalized acceptance condition, so that Spot would be usable with third-party tools that produce automata with various acceptance conditions. Doing so, we established a common format for exchanging \( \omega \)-automata. We have shown how we can make some acceptance transformation in order to realize certain operations (like emptiness checks) for which Spot does not yet support any acceptance conditions. Working on an emptiness check that works for general acceptance is definitely on our agenda; in fact this is a topic we have started to work on with colleagues from the Dresden University of Technology, and from the Masaryk University of Brno.

Chapter 5 briefly presented a SAT-based technique to minimize deterministic \( \omega \)-automata. This expensive procedure can be used to obtain minimal automata when that is desirable, and to discover areas where LTL translation and automata-simplification algorithms can be improved. We have also found this to be a useful tool to explore various generalized acceptance conditions, since we can now take a deterministic automaton with any acceptance and attempt to synthesize an equivalent automaton with another given acceptance condition. As these lines are written, an intern from EPITA is working towards making this minimization process more efficient using a better SAT-solver integration, some incremental SAT-solving techniques, and some new encoding ideas.

Finally in Chapter 6 we have presented some efficient ways to detect stutter invariant properties, and some results on using a generalization of what is known as testing automata. While the latter work (the subject of A. E. Ben Salem’s Ph.D. thesis) was done before we started working with generalized acceptance, it is now much easier to present using these notions, as has been done in Section 6.2. In fact, a complete generalization of testing automata would be to implement them as \( \omega \)-automata with arbitrary acceptance condition, but using changeset-labeled edges. It should be pointed out that the precise reason why the emptiness check using (generalized) testing automata is often better than using (generalized) Büchi automata is still unclear, and deserves further study.

7.2 Omitted Contributions

Four topics have been omitted from this document. The first two have not yet been published.
PSL, the Property Specification Language [Eisner and Fisman, 2006], is a huge industrial standard [PSL] that contains a linear fragment that is more expressive than LTL: it adds operators that allow using $\omega$-regular expressions to denote finite prefixes that should be followed by some temporal properties. Spot supports some of these PSL operators and the LTL translator has been extended to deal with the PSL operators and $\omega$-regular expressions in some non-trivial ways, and some specific PSL simplifications have been added. The supported PSL fragment is comparable to the Linear Dynamic Logic introduced by Vardi [2011]. Although the techniques we have implemented have been presented in two invited talks\(^1\), we have yet to write an article about it.

As mentioned on Section 4.6, Spot implements a determinization procedure that inputs transition-based Büchi automata and outputs automata with transition-based parity acceptance. The algorithm is based on the construction of Redziejowski [2012], augmented with some optimizations from ltl2dstar [Klein and Baier, 2006, 2007], as well as a few of our own. The most important improvement is to take the strongly connected components of the input automaton into account during the determinization. This would deserve at least a tool paper.

The next two topics have been published, but were omitted from this document for size reason.

In 2011, we studied two new “hybrid” model-checking approaches\(^2\)\(^3\), i.e., some techniques that are midway between explicit model checking (where the product $K_M \otimes A_{\neg \varphi}$ is represented as a graph) and symbolic model checking (where the product is represented symbolically, usually with decision diagrams). The two new techniques we presented, SOP (Symbolic Observation Product) and SLAP (Self-Loop Aggregation Product), both represent the product as a graph of aggregates: the nodes are symbolic sets of states of the original automata. Our experiment have shown that these techniques often outperform other existing hybrid or fully symbolic approaches.

Finally, we recently published a study of several provisos for Partial-Order Reductions.\(^4\) POR techniques are used during emptiness check to reduce the subset of $K_M$ that needs to be explored. They do so by locally ignoring some outgoing transitions whose corresponding action in the model $M$ is guaranteed to still be enabled later on. Ignoring such transitions can only be done if we guarantee that the corresponding action may not be continuously expanded along a cycle. A typical way to satisfy this constraint is to ensure that on each cycle, at least one state is fully expanded (i.e., none of its outgoing transitions are ignored). We reviewed several existing provisos to achieve this goal, and suggested new ones; discussing a total of 46 provisos.

\(^1\) At the Masaryk University of Brno in 2012, and at the Dresden University of Technology in 2015


8

Bibliography


M. Y. Vardi. The rise and fall of LTL. Invited talk to the 2nd International Symposium on Games, Automata, Logics and Formal Verification (GANDALF’11), 2011.

A

Selected uses of Spot

At the time of writing Google Scholar reports more than 126 citations of the paper in which we originally presented Spot in 2004\(^1\). This paper has remained the main paper to cite about Spot for 9 years, until we published some tool papers about Spot 1.0 in 2013\(^2\) (cited 21 times) and Spot 2.0 in 2016\(^3\). Not all these 147 citations are actual uses of Spot. In this appendix, we review a sample of the some work that have been built using Spot in some way.

A.1 Spot as an LTL-to-BA translator

Spot is used as an off-the-shelf translator from LTL to Büchi automata, or to generalized Büchi automata in context such as:

- creation of automata for monitoring C code [Staats and Heimdahl, 2008],
- generation of initial BA before SAT-based minimization [Ehlers and Finkbeiner, 2010],
- generation of automata before conversion to monitor [Tabakov and Vardi, 2010, Tabakov et al., 2012] before Spot could do it, or direct conversion to monitor [Bucur, 2012, Arcaini et al., 2013, Dutta and Vardi, 2014] now that it can,
- generation of DFA from LTL formulas that are guarantee properties [He et al., 2015],
- generation of BA before conversion to deterministic Rabin automata [Blahoudek et al., 2013, Sickert et al., 2016]
- generation of a GBA in a probabilistic model checker [Hahn et al., 2014] (they hope for a determinisic GBA, but will determinize the result to Rabin if not).

Spot has been used in several translation benchmarks, either by authors of other translators [Babiak et al., 2012, Mochizuki et al., 2014, Shan et al., 2014, e.g.], or by authors of benchmarks [Rozier and Vardi, 2007, Cichoń et al., 2009].

Some of those benchmarks are difficult to interpret because they do not always mention the version of Spot and the options used. Before Spot 1.0, there were no public command-line utility to translate LTL formulas into automata: Spot had a program called \texttt{ltl2tgba} in its test suite, which is what people used for benchmarking, but this


badly named binary was a crude front-end to most of the algorithms available in Spot, and it required a careful selection of several cryptic options to produce an optimized output. Today this testing program has been renamed, and there is a public tool called \texttt{ltl2tgba} that only does translation and has all optimizations turned on by default (favoring quality over execution time).

The benchmark of Rozier and Vardi [2007] and of Cichoń et al. [2009] have played some important role in the history of Spot, as we used them (after publication) to improve its performance. Some benchmarks released alongside Spot 0.9.1 show that the time needed to translate 100 formulas from Cichoń et al. [2009] was divided by 2.83 between version 0.8.3 and version 0.9.1.

Finally, Rozier and Vardi [2007] presents Spot 0.3 as the only tool of their benchmark that \textquote{can be considered an industrial quality tool} and consider Spot as \textquote{the best explicit LTL translator in our experiments}. Nonetheless using explicit translators (i.e., translator that creates automatas represented as graphs) for satisfiability checking is less efficient than using symbolic translators, so Spot is not the ideal tool to use for this task. Further work by Li et al. [2013] has built upon those two quotes to justify comparing a new LTL satisfiability solver against Spot; however, there exists more challenging tools that are actually dedicated to LTL satisfiability [Schuppan and Darmawan, 2011].

\subsection*{A.2 Spot as a research/development toolbox}

Tools distributed with Spot have many features that are convenient building-blocks for experiments. Here are some examples of uses:

- Generation of random LTL formulas [Molnár et al., 2015],
- Filtering of unique BAs up to one isomorphism [Blahoudek et al., 2016],
- Simplification of BAs [Blahoudek et al., 2016],
- Syntax conversion of LTL formulas [Maoz and Ringert, 2015],
- The \texttt{ltlcross} tool was used by the authors of \texttt{ltl3ba}, \texttt{ltl3dra}, \texttt{ltl2dstar}, and \texttt{Rabinizer3} to test recent version of their translators.\footnote{Personal communications with those tools’ authors.}

\subsection*{A.3 Spot as a library to build model checkers}

As Spot offers the bare minimum needed to build a model checker over a custom state-space, it was used in a couple of tools for this purpose.

MC-SOG [Klai and Poitrenaud, 2009] is a model-checker based on \textit{symbolic observation graphs} (SOG), in which each state symbolically encodes a set of states of the model. MC-SOG uses Spot for the dashed area of Figure 1.1 (p. 5), and implements the rest.

Neco, a compiler for turning Petri nets into native shared libraries that allow fast on-the-fly exploration of the state-space, was similarly linked with Spot to provide model-checking capabilities.\footnote{Ł. Fronc and A. Duret-Lutz. LTL model checking with Neco. In \textit{ATVA’13}, vol. 8172 of LNCS, pp. 451–454. Springer, 2013. Code moved to \url{https://github.com/Lvyn/neco-net-compiler}}
As already mentioned in Section 7.2 we generalized the SOG idea to Self-Loop Aggregation Product\(^7\) (SLAP) and Symbolic Observation Product\(^8\) (SOP). A model checking using these techniques has been implemented in ITS-Tools [Thierry-Mieg, 2015]. This time ITS-Tools provides an on-the-fly interface for building those hybrid products on-the-fly, and Spot is used over this interface to perform the emptiness check.

### A.4 Spot as an LTL or \(\omega\)-automaton library

Brotherston et al. [2012] used Spot in a theorem prover called Cyclist in which proof trees may be cyclic (hence the name) and need to satisfy some transition-based Büchi acceptance. Spot 0.8 was used since it was able to provide emptiness-checks for this type of acceptance, and also because it implemented a complementation algorithm (via a transition-based version of Safra’s determinization [Safra, 1989]). The current version of Cyclist now uses Spot 2 and its new transition-based determinization.

In the context of executing a workflow over multiple replicas for failure recovery, Schäfer et al. [2014] used the LTL translation routine of Spot to synthesize structurally different replicas from an identical LTL specification of the workflow. They modified the translation routine of Spot to include another parameter needed for their work.

Lemieux et al. [2015] use Spot just to parse LTL formulas and explore their syntactic tree.

Bauch [2015] and Barnat et al. [2016] used Spot as a C++ library for translating LTL formulas into automaton in a tool called Looney to check the sanity of real-world sets of requirements. They report that the translator they used before was inefficient on large formulas, and that switching to Spot “accelerated the process by several orders of magnitude on larger sets of formulae”.

Very recently, Blahoudek et al. constructed two tools using Spot:

- **ltl3hoa**\(^9\) translates LTL formulas into small non-deterministic automata using arbitrary acceptance conditions (as suggested by the LTL formula). It reuses the LTL parsing and simplification routines of Spot, as well as all the automata simplification routines; this way they focus only on their translation algorithm.

- **seminator**\(^10\) converts \(\omega\)-automata into semi-deterministic automata. They reuse the automaton parser of Spot, and its automata simplification routines. Again, they only needed to focus on the deterministic-algorithms they wanted to implement.

### A.5 Spot as a teaching environment

We use Spot at EPITA in an introductory lecture on model checking. Using the Python bindings in the IPython/Jupyter notebook, a web application for interactive programming [Pérez and Granger, 2007], the students can experiment with LTL and automata, and they can even build a small model checker by assembling the operations of Figure 1.1 (p. 5) in a few lines of Python code. Examples are given in Figure 2 of our ATV\(A’16\) tool paper.\(^{11}\)