Decoupling Synchronization from Logic for Efficient Symbolic Model Checking of Statecharts

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Abstract

Symbolic model checking is a powerful formal-verification technique for reactive systems. In this paper we address the problem of symbolic model checking for software specifications written as statecharts. We concentrate on how the synchronization of statecharts relates to the efficiency of model checking. We show that statecharts synchronized in an oblivious manner, such that the synchronization and the control logic are decoupled, tend to be easier for symbolic analysis. Thanks to this insight, the verification of some non-oblivious systems can be optimized by a simple, transparent modification to the model to separate the synchronization from the logic. The technique enabled the analysis of the statecharts model of a fault-tolerant electrical power distribution system developed by the Boeing Commercial Airplane Group. The results disclosed subtle modeling and logical flaws not found by simulation.

Keywords

Formal methods, verification, symbolic model checking, binary decision diagrams, requirements specification, statecharts, fault tolerance.

1 Introduction

Symbolic model checking [4] shows promise as an aid to producing industrial-strength software specifications in which developers have increased confidence [6, 19]. The formal languages for writing such specifications allow developers to produce specifications in a number of different styles. Just as the way that a program is written affects how efficiently one can analyze it, the style used to describe a specification affects how efficiently one can analyze it using symbolic model checking.

In this paper, we address how the synchronization in a statecharts specification—statecharts being one of the most broadly used languages for specifying reactive systems [9]—influences the efficiency of symbolic model checking. We identify certain styles for synchronization that are more efficient for symbolic analysis. For statecharts not written in these styles, we give procedures to automatically modify their internal representations to greatly improve the performance of their analysis.

This work started as a case study of applying symbolic model checking based on binary decision diagrams (BDDs) [3] to a statecharts specification developed by the Boeing Commercial Airplane Group. Previously, the same technique was applied to the requirements specification of the airborne collision avoidance system TCAS II [6, 7] written in the Requirements State Machine Language (RSML) [13], a language also based on statecharts. The observations and the optimization technique described in this paper result from the combined experience of these two case studies.

To elaborate, in symbolic model checking, the state space of a formal model is exhaustively explored. Sets of states are represented implicitly, so the method is not restricted by the state-space size and is able to analyze many systems much larger than conventional techniques can handle. The efficiency relies mainly on the succinctness of the symbolic representations such as BDDs, but their size is usually hard to predict. For software specifications, it depends not only on the functionality of the system but also on the particular way in which the specification is written.

Our model of statecharts responds to environment inputs by performing a macrostep, divided into a number of microstep transitions synchronized by events. We found that statecharts with events synchronized in an oblivious manner, such as the TCAS II requirements, tend to be more amenable to our analysis—the state machines’ synchronization is decoupled from their logic, resulting in fewer dependencies among the state variables and smaller BDDs. However, we observe that the length of a macrostep can often be bounded statically. By artificially incorporating a microstep counter into the state-
charts, we can decouple synchronization and logic in non-oblivious systems as well. The modifications are transparent to the specifier and preserve the model-checking results for most interesting properties (formally, all stutter-invariant properties, including all temporal-logic formulas without the next-time operator, are preserved [12]). The technique is interesting, particularly because it achieved substantial time and space improvements in our case study even though the numbers of state variables, reachable states, and search iterations were all increased, exactly the opposite of what most existing techniques attempt to do.

Another contribution of this work is the case study itself. Formal models have been used increasingly in Boeing to specify and validate functional requirements of airborne computing systems [15]. One of the modeling languages used is statecharts, thanks to their intuitive notations, ability to scale, and the availability of supported tools [16]. Developed for research purposes, the statecharts studied in this work model a fault-tolerant electrical power distribution (EPD) system designed for use on aircraft. Its purpose is to distribute electrical power from power sources to power busses via a number of circuit breakers, while tolerating failures in the power sources and circuit breakers. We were reasonably confident in the correctness of the model based on simulation results, but the model-checking analysis disclosed subtle modeling and logical flaws. Our efforts have been directed to finding bugs instead of verifying correctness. We give examples to argue for early use of model checking as a debugging tool because of the lower costs for analysis and the tendency of similar errors to recur in various parts of the system.

The rest of the paper is organized as follows. We first review statecharts and symbolic model checking in the next two sections. In Section 4 we explain the differences that oblivious and non-oblivious systems make to the efficiency of model checking. Our optimization technique is presented in Section 5. We describe the model of the EPD system and the results of the analysis in Section 6. Section 7 concludes the paper with some lessons learned.

2 Statecharts

Statecharts are a popular visual language for specifying complex reactive systems [9]. They extend state machine diagrams with parallelism, superstates, and broadcast communications. For simplicity, we will not discuss superstates in this paper (the techniques to be developed apply equally well to systems with superstates). Instead, our system model consists of a finite set of parallel local state machines, with a finite set of events and inputs, all embedded in a nondeterministic environment.

2.1 Syntax and Semantics

Figure 1(a) gives a simple example with two parallel state machines A and B, synchronized by events x, y, and z. Arrows without sources indicate the initial local states. Other arrows represent local transitions, which have labels of the form trig(\text{cond})/acts, where trig is a trigger event, cond a guarding condition, and acts a (possibly empty) list of action events. The guarding condition is a predicate on local states of other state machines and/or inputs to the system. A label of the form trig[true]/acts is abbreviated as trig/acts. The transition is enabled whenever the event trig occurs and the guarding condition cond evaluates to true.

An external event is one that can be generated by the environment. For simplicity, we assume that an external event cannot also be an action event of any transition, and call events that are not external internal. Initially the machines are in their respective initial local states, and the environment generates a subset of the external events and arbitrarily changes the inputs to the system, enabling transitions as described above. Different statechart-based languages disagree on which enabled transitions are taken and what effects the taken transitions produce. We adopt the semantics of RSML [13] and STATEMATE [10]: Two transitions are non-conflicting if they do not share the same source local state, and a maximal set of enabled transitions that are pairwise non-conflicting, collectively called a microstep, is simultaneously taken—the system leaves the source local states of the transitions, enters the destination local states, and generates the action events (if any). The generated action events may trigger additional transitions in the next microstep. Events are instantaneous, so unless regenerated they disappear after the microstep.

In our example in Figure 1(a), both machines are initially in off. Event x is assumed to be external. The guarding condition c is assumed to be a Boolean input, but it could have

![Figure 1](image-url)

Figure 1: Two ways to specify a controller (A) and a plant (B) in statecharts, with equivalent stable-state behaviors
been a predicate on other local machines not shown. When \( x \) occurs and \( c \) is true, the transition from \( \text{off} \) to \( \text{on} \) in machine \( A \) is enabled and taken, generating \( z \). The event \( z \) in turn triggers the transition from \( \text{off} \) to \( \text{on} \) in machine \( B \) in the next microstep, and generates \( v \). In general, whenever \( x \) occurs, machine \( A \) (the controller) will be in \( \text{on} \) if and only if \( c \) is currently true; machine \( B \) (the plant) follows and lags behind machine \( A \) for one microstep.

The system is \textit{stable} when no events occur. The sequence of microsteps between the time when the system is unstable and the time when it becomes stable again is called a \textit{macrostep}. (A microstep and a macrostep are called a step and a superstep respectively in STATEMATE, whereas in RSML they are called a microstep and a step respectively.) The \textit{synchrony hypothesis} says that during a macrostep no external events can arrive and the environmental inputs remain unchanged; that is, the system is infinitely faster than the environment [1]. Figure 2 depicts these notions. RSML enforces the synchrony hypothesis, while STATEMATE optionally allows it. We assume the synchrony hypothesis, which is central to the issues and techniques discussed in this paper.

### 2.2 Styles: Oblivious vs. Non-Oblivious

Figure 1(b) shows another way to specify the controller and plant. Instead of generating events \( y \) and \( z \) to turn on and off machine \( B \), machine \( A \) now generates event \( w \) to signal its completion and pass the execution to machine \( B \), which reacts based on the \( A \)'s local state. We call such statecharts \textit{oblivious} in the sense that the sequence of events generated and thus the synchronization are independent of the local states or inputs; in this case, \( w \) is generated after \( x \) regardless of the condition \( c \) and the local state of \( A \). More explicitly, a difference between the two systems arises when external event \( x \) occurs but, say, machine \( A \) is \text{off} and \( c \) is false, in which case in Figure 1(a) no transitions are enabled and no internal events generated. In the same situation, \( w \) is still generated in Figure 1(b). Despite the difference, the stable-state behaviors of the two systems are identical.

A few observations are worth noting. In the non-oblivious system, the events are used for both synchronization (executing machine \( B \) after machine \( A \)) and logic (directing machine \( B \) to the appropriate local state), and the specifiers is more concerned about the local, microstep-level interaction between the two machines. In contrast, in the oblivious system, events are merely used for synchronization—the logic is specified in the local states and the guarding conditions, and the specifier foresees the overall control flow between the machines in a macrostep and constructs events to synchronize the machines in the desired order. Oblivious systems thus have fewer dependencies, as an event depends upon nothing but other events. While virtually all of the STATEMATE machines that we have seen are not oblivious, the portion of the RSML specification of TCAS II that we analyzed (in fact most of the entire specification) is oblivious. This is consistent with Harel and Naamad’s comment that in RSML a macrostep appears to be the “basic operation,” while in STATEMATE a microstep is the basic operation [10, p. 323]. Notice, however, that the differences arise not from the syntax or semantics, but from the distinct mental models of the system that the specifiers have.

### 3 Symbolic Model Checking

We review the model-checking problem and the idea of symbolic model checking in this section.

#### 3.1 The Model-Checking Problem

To analyze statecharts using state-exploration techniques, we view the system as a \textit{global structure} \( (Q, R, I) \), where \( Q \) is a finite set of (global) states, \( R \subseteq Q \times Q \) a total transition relation, and \( I \subseteq Q \) a set of initial (global) states. A state in \( Q \) is a tuple of the current local state of each state machine, the set of events occurring, and the values of the environmental inputs. A \textit{path} is an infinite sequence of states in which each consecutive pair of states is in \( R \), and a \textit{trace} is a path that starts with some initial state in \( I \). A state is \textit{reachable} if it appears on some trace.

We symbolically encode the state space \( Q \) by declaring a set \( V \) of state variables as follows. For each state machine, declare a state variable whose range is the local states of the machine. For each event, declare a Boolean state variable, which is true if and only if the event occurs. For each input, declare a state variable with the same range (assumed finite). Clearly, this mapping from \( Q \) to the valuations of the state variables in \( V \) is one-to-one. We will not distinguish between a state variable and its encoded statechart entity (local state, event, or input) because of their simple correspondence.

Given this encoding, the set of initial states \( I \) is represented as \( \forall m \in M \, m = m_0 \land \forall e \in E, \neg e \), where \( M \) is the set of state machines, \( m_0 \) is the initial local state of \( m \), and \( E \) is the set of internal events. This simply says that initially, each machine is in its initial local state; all the internal events do not occur, but the external events and inputs are not constrained. More interesting is the encoding of the transition relation \( R \) [6]. It suffices here to point out that our encoding has the form \( \neg \text{stable} \rightarrow \text{micro} \land (\text{stable} \rightarrow \text{env}) \), where \( \text{micro} \) encodes microsteps, \( \text{env} \) encodes the environmental transitions across macrosteps, and \( \text{stable} \) indicates when the system is stable, namely \( \forall e \in E, e \) where \( E \) is the set of all events. Note that under this encoding a macrostep is represented as a sequence of global transitions. An alternative is to represent a macrostep
as a global transition, but this would prevent us from analyzing behaviors within a macrostep, which is often useful for debugging purposes.

Many system properties can be expressed in the Computation Tree Logic (CTL) [8], a common temporal logic for model checking. Its formulas are built from propositions (predicates over the state variables in $V$), the usual Boolean operators, path quantifiers $A$ (for all paths) and $E$ (for some path), and modalities $X$ (next-time), $F$ (eventually), $G$ (always), $U$ (strong until), and $W$ (weak until), with every modality immediately preceded by a path quantifier. Intuitively, each modality is evaluated over a path, and $X\phi$ means that $\phi$ holds on the path starting at the next state, $F\phi$ means that $\phi$ holds somewhere on the path, $G\phi$ means that $\phi$ holds everywhere on the path, $\phi U \psi$ means that $\psi$ holds somewhere on the path and $\phi$ holds everywhere before that, and $\phi W \psi$ means that $\phi$ holds everywhere before $\psi$ holds, but if $\psi$ never holds, then $\phi$ must hold forever. For example, the formula $AG\text{safe}$ asserts that the proposition $\text{safe}$ holds in all reachable states, $AG(request \rightarrow A\text{Finished})$ asserts that a request always results in a response in the future, and $AG(request \rightarrow A(requestW\text{response}))$ asserts that once issued, a request will persist unless a response is given.

Given a global structure and a temporal-logic formula, the model-checking problem asks whether the structure satisfies the formula. If not, to provide valuable diagnostic information, a model checker usually gives a counterexample, a trace that falsifies the property.

### 3.2 Symbolic Search

The truth value of a formula can be found by searching the state space. We define $\text{Pred}$ and $\text{Succe}$: $2^V \rightarrow 2^Q$ to compute respectively the predecessors (also called pre-image) and successors (also called image) of a set of states under the transition relation $R$:

$$\text{Pred}(S) = \{ q \in Q \mid \exists q' \in S. (q, q') \in R \}$$

$$\text{Succe}(S) = \{ q' \in Q \mid \exists q \in S. (q, q') \in R \}.$$

Consider the formula $Ag_p$, which is true if and only if the proposition $p$ holds in every reachable state. Let $P \subseteq Q$ be the set of states that satisfies $p$. As shown in Figure 3, we can evaluate the formula by performing either a forward breadth-first search from the initial states $I$ to find the set $Z$ of reachable states, or a backward breadth-first search from the set $Y_0$

<table>
<thead>
<tr>
<th>Forward:</th>
<th>Backward:</th>
</tr>
</thead>
<tbody>
<tr>
<td>$Z_0 := I; Z := Z_0; i := 0$</td>
<td>$Y_0 := Q - P; Y := Y_0; i := 0$</td>
</tr>
<tr>
<td>repeat $i := i + 1$</td>
<td>repeat $i := i + 1$</td>
</tr>
<tr>
<td>$Z := Z \cup Z_i$</td>
<td>$Y := Y \cup Y_i$</td>
</tr>
<tr>
<td>$Z_i := \text{Succe}(Z_{i-1}) - Z$</td>
<td>$Y_i := \text{Pred}(Y_{i-1}) - Y$</td>
</tr>
<tr>
<td>until $Z_i = \emptyset$</td>
<td>until $Y_i = \emptyset$</td>
</tr>
</tbody>
</table>

$Ag_p$ iff $Z \subseteq P$. $Ag_p$ iff $Y \cap I = \emptyset$.

Figure 3: Forward and backward searches for $Ag_p$ of states that immediately violate $p$ to find the set $Y$ of states that may eventually violate $p$. The loops, guaranteed to terminate for finite state spaces, are said to compute fixed points. In conventional "explicit" search, $Z$ and $Y$ are implemented as hash tables, while the search frontiers $Z_i$ and $Y_i$ are implemented as queues. More complicated temporal-logic formulas can be evaluated in similar ways by computing one or more fixed points [8].

The method is impractical for many large systems because of the sheer number of states that must be explored. More efficient for large state spaces are symbolic searches [4]. A state set (e.g., $Z_i$) can be symbolically encoded as a predicate over the state variables, just as we encoded the initial states $I$ in Section 3.1. The idea then is to manipulate this predicate directly to explore the whole set without enumerating its elements. Because we are dealing with finite state spaces, we can assume without loss of generality that each state variable is Boolean, so each such predicate is a Boolean function, which can be represented as reduced ordered binary decision diagrams (BDDs) [3]. Boolean operations, satisfiability checking, and successor and predecessor computations can be performed efficiently using BDDs, which therefore can be used to implement the forward and backward searches described above. BDDs are canonical, meaning that each Boolean function has a unique BDD representation up to a chosen variable order. In addition, a variable never appears in the BDD if the function does not depend on the variable.

The size of the BDDs is a major bottleneck in BDD-based algorithms. In the worst case it can be exponential in the number of variables. In practice though, it is often small even when the set represented is large, but this depends on the chosen variable order and the dependencies among the variables.

### 4 Some Intuitions on BDD Size

In this section, we examine how the BDD size can be affected by the style of synchronization. Intuitively, the decoupled synchronization and logic of oblivious systems induces fewer dependencies among the state variables, often keeping the BDDs smaller. For instance, non-oblivious systems have many more ways to finish a macrostep than oblivious ones, and a backward search from the stable states needs to capture all these possibilities, resulting in larger BDDs. The difference for forward searches, however, is less apparent. Forward searches from the initial states maintain all system invariants, which are determined mostly by the functionality of the system. Whether or not the synchronization is oblivious, there may be many such invariants relating local states, events, and inputs in nontrivial ways, and these may result in very large BDDs [11]. As a consequence, we have not found forward searches efficient in our experiments, so we focus on backward searches in this paper.

For an example that illustrates the intuition on backward searches, consider the non-oblivious system in Figure 1(a). We encode the state space with Boolean state variables $x, y,$
z, v, a, b, and c, where a (resp. b) is true if and only if machine A (resp. B) is in on. Refer to the backward search in Figure 3. Assuming that we want to know whether machine A can be in on in a stable state, that is, \( Y_0 = \text{stable} \land a = \neg(x \lor y \lor z \lor v) \land a \), what are its predecessors in \( Y_1 \)? Such a state must be unstable, so one of the events must occur. In addition, in order for the next state to be stable, none of the transitions should be enabled. So to be in on in the next state, machine A must be in on already. Combining all these conditions, symbolically we have

\[
Y_1 = (x \lor y \lor z \lor v) \land a \\
\land (x \rightarrow c) \land (y \rightarrow \neg b) \land (z \rightarrow b),
\]

quite complicated for even this simple example. The complexity stems from the numerous ways in which a macrostep may terminate; some of them are in fact not possible (for example, the conjunct \( y \rightarrow \neg b \) could have been simplified to \( \neg y \), because \( y \land \neg b \) is not reachable).

Now consider the oblivious system in Figure 1(b). We have \( Y_0 = \text{stable} \land a = \neg(x \lor w \lor v) \land a \). Unlike the situation above, however, a predecessor of \( Y_0 \) cannot have events \( x \) or \( w \) occurring, because otherwise \( w \) or \( v \) respectively will be generated, leading to an unstable state. So we have \( Y_1 = v \land a \), much simpler than the previous case. In the subsequent iterations, we have \( Y_1 = w \land a \), and \( Y_2 = x \land c \), etc., all simple expressions. In addition, unlike the non-oblivious case, the local states of \( B \) never appear in the expressions and therefore the BDDs. This is appealing because the property being checked indeed does not depend on \( B \). Ultimately, the fixed points computed in both cases are the same, but the BDDs for the search frontiers are smaller for the oblivious system.

The example was designed to be simple just to give us some intuition. The BDDs that occur in actual analysis are much more complex, but our experimental results so far are consistent with our reasoning that backward searches tend to be much more efficient for oblivious than non-oblivious systems.

5 Optimization

In the previous section, we saw some reasons why non-oblivious machines can be hard for symbolic backward searches. Armed with that intuition, we systematically modify the global structure to decouple the synchronization from the logic of the system while preserving most system properties: Often, the maximum length of a macrostep can be statically bounded by analyzing the dependencies among the events. In this case, we can make every macrostep equal in length, and incorporate a microstep counter into the system. This counter is oblivious in that its behaviors do not depend on the internal events or the state machines, and is used to guard every local transition.

5.1 Microstep counter

We need several definitions. For each \( e_1 \) and \( e_2 \) in the finite set \( E \) of events, we say that \( e_1 \) precedes \( e_2 \), written \( e_1 \prec e_2 \), if there exists a transition labeled with \( e_1[c]/e_2 \) for some guarding condition \( c \). This precedence relation \( \prec \) is assumed to be acyclic, that is, \( (e, e) \notin \prec \) for each \( e \), where \( \prec \) is the transitive closure of \( \prec \). Many systems have this property because it prevents the nontermination of macrosteps, a design flaw that is potentially hard to locate. (This assumption is not essential for our technique, but it makes bounding the maximum length of a macrostep easy, as we now show.)

For each event \( e \), let \( \lambda(e) \) be the smallest set of integers such that

1. \( \lambda(e) = \{1\} \) if \( e \) is an external event,
2. \( i \in \lambda(e) \) implies \( i + 1 \in \lambda(e') \) for all \( e' \) with \( e \prec e' \).

Intuitively, \( i \) is in \( \lambda(e) \) if \( e \) can occur just before the \( i \)th microstep of some macrostep. Since \( \prec \) is acyclic, the integers in \( \lambda(e) \) are bounded, and the values of \( \lambda(e) \) for all \( e \) can be computed in \( O(|E|^3) \) total time by traversing the event-precedence graph. The maximum length \( d \) of a macrostep is then the largest integer in \( \lambda(e) \) for any \( e \). For Figure 1(a), we have \( x \prec y, x \prec z, y \prec v, z \prec v \), and thus \( \lambda(x) = \{1\} \), \( \lambda(y) = \lambda(z) = \{2\} \), \( \lambda(v) = \{3\} \), and \( d = 3 \). Note that some macrosteps may have fewer than \( d \) microsteps.

To symbolically encode a statecharts model as a global structure, in addition to the usual state variables, we define a microstep counter \( mc \) to range from 0 to \( d \). The behavior of the microstep counter depends only on the set \( E_s \) of external events (the primed variables below encode the next state):

**Modification 1 (Microstep counter).** Let \( s \) denote \( \bigvee_{e \in E_s} e \) (some external event occurs in the current state) and \( s' \) denote \( \bigvee_{e \in E_s} e' \) (some external event occurs in the next state). We conjoin the symbolic encoding of the initial states \( I \) with

\[
(-s \rightarrow mc = 0) \land (s \rightarrow mc = 1),
\]

and conjoin the transition relation \( R \) with

\[
((mc = 0 \land \neg s') \rightarrow mc' = 0) \land ((mc = 0 \land s') \rightarrow mc' = 1) \land (0 < mc < d \rightarrow mc' = mc + 1) \land (mc = d \rightarrow mc' = 0).
\]

Stability now depends only on the microstep counter:

**Modification 2 (Stability).** The proposition \( \text{stable} \) is now defined as \( mc = 0 \).

The rules intuitively say the following: If no external event occurs in the initial state, the system is stable and \( mc \) is initialized to 0. Whenever some external event occurs, \( mc \) becomes 1 in the same state and a macrostep begins. The value of \( mc \) is then incremented by one in every subsequent microstep until the value reaches \( d \). At that point, it is reset to 0 in the next state and the system is considered stable. Note that the internal events do not come into the picture, and that every macrostep has exactly \( d \) microsteps.
Clearly, the local transitions in the statecharts are unaffected by the changes, but the stable state may be delayed as illustrated in Figure 4—when the original system is stable, the modified system may still be incrementing \( mc \). However, because the microstep counter is not visible to the user, the modified system will not produce any visible change until stable. Formally, the system stutters in the interim [12], and every CTL formula without the next-time \( X \) operator is preserved by stuttering [2]. (Intuitively, formulas with the \( X \) operator can count the number of microsteps and thus may not be preserved.)

Our final modification uses the microstep counter to guard transitions.

**Modification 3 (Guards).** Each transition labeled with \( e_1[\text{cond}]/e_2 \) is encoded in the global structure as if it were labeled with \( e_1[\text{cond} \land mc \in \lambda(e_1)]/e_2 \).

One can intuitively think of the new label as

\[
mc \in \lambda(e_1)[e_1 \land \text{cond}] / e_2.
\]

In other words, the transition is triggered by the microstep counter, and the event \( e_1 \) becomes part of the logic of the guarding condition. Notice, however, that this modification cannot affect the system’s behavior, because in any reachable state, the occurrence of \( e_1 \) implies \( mc \in \lambda(e_1) \). This can be proved by induction on the definition of \( \lambda \). So the inclusion of \( mc \in \lambda(e_1) \) is redundant as far as forward behavior is concerned. We make the following claim:

**Claim (Correctness).** If the event-precedence relation \( \prec \) is acyclic, then Modifications 1–3 preserve every CTL formula that does not contain the \( X \) operator and does not refer to the value of the microstep counter (except in indirectly comparing it with zero by referencing \( \text{stable} \)).

To see how these modifications help backward searches, consider again the example in Section 4: We want to search backward in the non-oblivious system from \( Y_0 = \text{stable} \land a \). Figure 5 shows the modified machines with transitions guarded by the microstep counter. Recall that \( \text{stable} \) means \( mc = 0 \), and by the construction of \( mc \), an unstable predecessor of a stable state must have \( mc = d = 3 \), implying that the transitions in \( A \) cannot be enabled. So to be in \( \text{on} \) in the next state, \( A \) must be in \( \text{on} \) already. That is, \( Y_1 \) is \( mc = 3 \land a \), a lot simpler than the search for the non-oblivious system in Section 4. In the subsequent iterations, \( Y_2 \) will be \( mc = 2 \land a \), and \( Y_3 \) will be \( mc = 1 \land c \). In fact, in this example, the search looks a lot like the one for the oblivious system.

In fact, the modifications improve the efficiency in another way, which also helps analyze oblivious systems. Not evident in our simple examples, a source of inefficiency common to both types of systems is the exploration of unreachable simultaneous transitions, because a backward search does not know that, for instance, in Figure 1(a) the transitions in machine \( A \) can never be enabled simultaneously with any transition in machine \( B \) in any reachable state. The microstep counter in Figure 5, however, makes this fact explicit and helps prune such simultaneous transitions in the search. Previously, this problem was tackled by identifying a set of events that are always mutually exclusive [7]. The method presented here is more general and more effective.

Recall that our construction of the microstep counter makes certain macrosteps longer so as to make every macrostep equal in length. This generally results in an increased number of iterations to reach the fixed point in the search, affecting the performance in a negative way. Nevertheless, in our case study reported in Section 6, this impact was negligible compared with the benefits of reducing the BDD size. The lengthened macrosteps also introduce extra states in a counterexample, but these states are easy to detect and can be removed to recover the actual counterexample.

### 5.2 Condition-Driven Transitions

Some variants of statecharts, such as STATEMATE, allow transitions not guarded by events. That is, a transition can have labels of the form \( [\text{cond}]/\text{acts} \), where \( \text{cond} \) is a guarding condition and \( \text{acts} \) is a list of action events. Such transitions are enabled when the machine is in the source local state and \( \text{cond} \) is true. We call these transitions **condition-driven**. Intuitively, instead of checking the guarding condition only when being triggered by an event, condition-driven transitions continuously poll the guarding condition.

Extending our techniques to handle condition-driven transitions requires a more general framework, and we omit the details in this paper. The basic idea, though, remains the same. Specifically, when we encode a transition \( t \), we want to conjoin its guarding condition with a new proposition \( mc \in \rho(t) \) where \( \rho(t) \) is a set that includes an integer \( i \) if transition \( t \) can be taken in the \( i \)th microstep. For systems with every transition triggered by some event, the set \( \rho(t) \) is simply \( \lambda(e) \) with \( e \) being the trigger event of \( t \), as we saw in Modification 3.

In the presence of condition-driven transitions, we can still com-
puter $p$ statically in many common cases, although the procedures are more involved.

6 Case Study

The techniques presented in the previous section were motivated by the analysis of a statecharts model of the electrical power distribution (EPD) system on the Boeing 777 aircraft. We briefly describe the model, discuss some results of the analysis, and report the benefits of the optimization techniques given in the previous section. We also discuss how model checking could potentially be used to benefit the model-based development processes used at Boeing. We stress that the statecharts model was developed for research purposes and does not represent the actual requirements used to develop the on-board system. As such the model by intent did not include all the logic necessary for a complete specification. The model was intended as a high-level abstraction of the electrical system, which included only the logic necessary to accomplish the goals of a wider airplane system analysis [16].

6.1 The EPD Model

The purpose of the EPD system is to distribute AC and DC power to other airplane systems. It comprises separate interconnected distribution systems including main AC power, backup AC power, DC power, standby power and flight controls power. Electrical power is distributed from power sources to power busses via a number of relayed circuit breakers. Failures of the power sources or circuit breakers are automatically detected and isolated. We focus on the portion of the statecharts that models the main and backup AC distribution subsystems. There are 33 two-state machines, 23 Boolean inputs, and 34 events, for a total of 90 Boolean state variables, or about $10^{27}$ global states, of which at least $10^{15}$ are reachable.

Figure 6(a) depicts part of the system configuration in normal operations. The power busses $l_{\text{main}}$ and $r_{\text{main}}$ belong to the main AC power subsystem, and are normally powered by the generators $l_{\text{gen}}$ and $r_{\text{gen}}$ respectively. When $l_{\text{gen}}$ loses its power because of either manual shutdown or failure, the circuit breakers will be reconfigured automatically to use $r_{\text{gen}}$ to power both $l_{\text{main}}$ and $r_{\text{main}}$, as illustrated in Figure 6(b). The same configuration may also result from failures in the circuit breakers that connect $l_{\text{gen}}$ and $l_{\text{main}}$. The

![Figure 7: A circuit breaker (CB) and its controller (CTRL) system is supposed to satisfy a number of stringent requirements, such as the resilience of the power busses against single or multiple failures in the power sources and/or the circuit breakers.](image)

A circuit breaker, either open or closed at any time, is modeled as a two-state machine and is managed by a controller. Figure 7 shows a generic circuit breaker and its controller. The transitions in the circuit-breaker state machine are guarded by the complement of a Boolean input $f$ that indicates a failure, so a failed circuit breaker does not respond to the controller. The guarding condition $c$ of the controller is usually a nontrivial predicate relating inputs, the local states of other circuit breakers, as well as the power sources.

6.2 Analysis

We analyzed the main and backup AC power subsystems by translating the statecharts to the input language of the CTL model checker SMV [14]; other subsystems were abstracted away manually. The analysis can be divided into analysis on normal behaviors (i.e., no component failures) and fault tolerance (single and multiple failures). We report some of the more interesting results here. Although the model had been exercised extensively in simulation, several flaws were discovered using model checking. We were able to obtain these results only after using our optimization technique presented in Section 5. Performance data will be given later.

Normal Operations

In normal operations, all busses in the main and backup AC subsystems should be powered in the stable states. We checked the formula

$$\text{AG}(\text{stable } \land \neg \text{no-failures}) \rightarrow (\text{main } \land \text{backup})$$  \hspace{1cm} (1)$$

where $\neg \text{no-failures}$ is a proposition indicating the absence of failures, and $\text{main}$ and $\text{backup}$ assert respectively that the main busses ($l_{\text{main}}$ and $r_{\text{main}}$) and backup busses are powered. Note that the formula does not simply ignore failures; it takes into account scenarios in which failures occur but are subsequently recovered. The formula was evaluated true by the model checker.

Not only should the busses be powered when there are no failures, they should be powered by different sources. We checked the formula

$$\text{AG}(\text{stable } \land \neg \text{no-failures}) \rightarrow \text{separate-sources}$$  \hspace{1cm} (2)$$
where the proposition separate-sources asserts that a power source is connected to at most one bus. This time, however, the model checker gave a counterexample revealing a bug in the model of the circuit breakers. In the counterexample, \( r_{gen} \) initially powers both \( l_{main} \) and \( r_{main} \) because of a failure in the circuit breakers. Now assume the failed circuit breaker is modeled by the machine \( CB \) in Figure 7. The recovery of \( CB \) corresponds to the Boolean input \( f \) changing to false. This change alone, however, cannot trigger any local transition, as the transitions in \( CB \) are guarded by events. So when \( CB \) recovers, the system ends up in a situation in which there are no failures, but \( r_{gen} \) is still powering both main busses, violating the property. We refer to this bug as B1, which we fixed by making \( CB \) go to the local state indicated by its controller upon recovery. With this bug fix, the formula was successfully verified.

**Fault Tolerance**

The main busses should in fact tolerate one failure in the power sources or circuit breakers. We checked the formula

\[
\text{AG}( (\text{stable} \land \text{at-most-1-failure}) \rightarrow \text{main}) \quad (3)
\]

where the proposition at-most-1-failure has the obvious meaning. The model checker gave a counterexample that again reveals the bug B1, although the scenario is more complex. It involves a failure in a circuit breaker, a change in inputs to induce a state change in its controller, the circuit breaker’s recovery, and a subsequent failure in one of the power sources. After we fixed the bug and rechecked the property, the model checker gave another counterexample that discloses a logical flaw—one of the circuit breakers does not respond to a failure in another circuit breaker that it is supposed to handle, resulting in power loss to both main busses.

We refer this bug as B2. (We have not attempted to fix this bug in this study.)

The backup busses should also tolerate one failure. We fixed B1 in all the circuit breakers and successfully verified the analog of Formula 3 for the backup busses. We initially thought that the backup busses should survive two failures. We checked this stronger property to which the model checker gave a counterexample exposing a logical flaw similar to B2 above. The counterexample involves simultaneous failures of two power sources, their subsequent recovery, and then simultaneous failures of two circuit breakers.

**Miscellaneous**

The formulas above are only concerned about stable states. One might expect certain causality to be maintained even in the unstable states. For example, the formulas do not prevent, within a macrostep, the power from going off before failures occur, as long as the right thing happens at the end of the macrostep. So, we evaluated formulas such as

\[
\text{AG}(\text{main} \rightarrow \text{A(main W-no-failures)}) \quad (5)
\]

which asserts that, even in the unstable states, if the main busses are powered, then the power should persist unless a failure occurs. Interestingly, the model checker showed various scenarios violating such properties—some situations that we do not regard as failures can cause transient power loss to the busses. Although this does not reflect any flaw in the system, it is still an interesting find as the scenarios were not obvious to us before the analysis. Such results can provide insights into the design of the model and can reveal design flaws in some cases.

Other properties that we verified include the impossibility of having certain circuit breakers closed simultaneously (which would indicate some illegal system configuration), and other sanity checks, such as the property that if no power sources are operating, then no busses should be powered.

### 6.3 Performance

Despite the results we finally obtained, the initial experience of the analysis was daunting—the BDDs generated were enormous and the fixed-point computations could not go beyond two or three iterations before we ran out of memory. As a result, even trivial properties could not be analyzed, let alone the formulas given above. We attacked the problem by focussing on a small part of the system, trying alternative ways of modeling, and looking for specific reasons for the BDD blowup. Hand-simulation of the symbolic search was sometimes used to build up intuition. The optimization technique given in Section 5 results from the insights gained in the process.

Table 1 shows, for each property, the number of search iterations, the time (in seconds), and the number of BDD nodes (in thousands) needed to compute the fixed point. (Formula 5 requires computing four separate fixed points to evaluate, and its number of iterations in the table is the sum of the four numbers.) All these searches were performed on the model without fixing the bugs B1 or B2. The results were obtained on a Sun Ultra-2 workstation using SMV version 2.4.4 (augmented with a conjunctive partitioning heuristic [17]). The data suggest dramatic improvements made by our optimization technique, without which the evaluation of each of the properties was not feasible. The rightmost column gives the data for computing the reachable states using a forward search, and shows the superiority of backward searches for our system.

We add that fixing B1 in the optimized model dramatically reduces the time taken to evaluate each property to less than ten
searching could potentially help to ensure that the model reflects other key design goals in that many of the system properties discovered during model checking might have been catalogued critically on our optimization technique, which aims at reducing the size of the BDDs representing state sets. In hardware verification, techniques with the same goal exist and usually work by altering these BDDs dynamically during the search [5, 11, 18]. They are quite general and work for large classes of circuits. We applied some of these techniques to the EPD model, but the results were not satisfactory. In contrast, the technique we developed concentrates on statecharts and statically changes the underlying global structure. Intuitively, the technique uses information from forward analysis on event precedence to realign the search frontiers and to prune backward searches. This strategy of combining forward syntactic analysis and backward searches appears to be a promising approach to improving the efficiency of symbolic model checking.

A major goal of the case study was to evaluate the use of model checking as a debugger in support of requirements validation at Boeing by providing an additional debugging tool over and above the existing use of simulation. The use of modeling and simulation to support requirements validation at Boeing is described in [15]. In this process, the written specification is developed first, and then a model is created to assist in validation of the requirements. Typically the model is simulated and executed by providing user-oriented inputs to the model and monitoring responses through panel graphics that represent actual system interfaces. Model checking could potentially help to ensure that the model reflects other key design goals in that many of the system properties checked in this case study are not revealed in the operator interface.

Some flaws found during model checking might have been found if simulation runs had been explicitly defined to test conformance. However, the simulations would have had to include an extensive test suite, which included cases of intermittent failures of components to find the class of errors found during our model checking. Model checking appears to be particularly beneficial in helping find these “corner cases” with a minimum of additional effort.

The analysis described was done several years after the development of the model. However, it is clear to us that use of model checking during the initial development of the model would have detected subtle flaws in logic before they were repeated throughout a much larger model. For example, the bug B1 repeats in every state machine that models a circuit breaker, and bugs similar to B2 appear in several parts of the model. In fact, some of these flaws could be found by focusing on the main AC subsystem and ignoring the backup AC subsystem.

7 Conclusion

We have made several contributions in this work. We carried out a case study of applying BDD-based model checking to a statecharts specification developed at Boeing and discovered subtle flaws in the model. The experience enabled us to identify oblivious synchronization as a feature of statecharts that facilitates model checking—the decoupled synchronization and control logic tend to make the BDDs representing state sets smaller when backward symbolic search is used. We devised an innovative technique of introducing a microstep counter into the model to achieve a similar effect in systems with non-oblivious synchronization. The technique works by statically bounding the length of a macrostep and using the microstep counter to synchronize local transitions. The improvements were crucial for the case study, as they allowed analysis that used to be infeasible to complete in just several minutes of CPU time.

Getting intuition on BDD size in general is notoriously hard, because the size does not directly correlate to simple measures such as the number of variables or the number of reachable states. However, formal software specifications are often written in a few common styles or using a few popular idioms, and it may be possible to gain enough insights to optimize for these common cases. This work follows this direction and contributes to a better understanding of the tradeoffs between specification and verification. We hope that the results will be valuable for designing specifications or specification languages that are more amenable to symbolic model checking.

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References


