# Simulation of Simultaneous Events in Regular Expressions for Run-Time Verification

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#### Abstract

When specifying system requirements, we want a language that can express the requirements in the simplest and most intuitive form. Although the MaC system provides an expressive language, called MEDL, it is generally awkward to express certain features like temporal ordering of complex events, timing constraints, and frequencies of events which are inherent in safety properties. MEDL-RE extends the MEDL language to include regular expressions to easily specify timing dependencies and timing constraints. Due to simultaneous events generated by the MaC system, monitoring regular expressions by simulating DFAs would result in a potential problem. The DFA simulations would involve concurrent multi-path simulations and result in exponential running time. To handle simultaneous events inexpensively, we generate a dependency graph to identify possible simultaneous events. Further, we augment the original DFAs with alternative transitions, which will substitute for multi-path simulations.

## 1 Introduction

The monitoring, checking and steering (MaC) framework [9,10,11] has been designed to ensure that the execution of a real-time system is consistent with its requirements at run-time. It provides a language, called MEDL, to specify safety properties based on LTL [13]. The safety properties include both computational and timing requirements. The safety properties are defined in terms of events, conditions, auxiliary variables, and auxiliary functions. Events are instantaneous incidents such as variable updates or the start/end of a method call. Conditions are propositions about the program that may be true or false for a duration of time. Those events and conditions can also

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be composed using connectives described in Section 2. Auxiliary variables are temporary storage, which allows us, for example, to count the number of occurrences of an event. Auxiliary functions return values and time stamps of events. The MEDL language provides an elegant and intuitive way to specify computational requirements. It, however, does not provide an intuitive way to specify timing requirements, such as temporal ordering of events with complex timing dependencies, timing constraints or counting of specific events in a time interval.

The extension of MEDL called MEDL-RE [14] adds the ability to specify ordering of events in the form of regular expressions (RE) over a customized set of events, which offers users with clearer and less error-prone specifications. In this paper, we propose an efficient simulation of the corresponding DFAs at runtime. By observing a sequence of events occurring in a target system, a DFA generated by MEDL-RE matches the sequence of events with a specific regular expression. However, because the composite events can be triggered simultaneously and cannot be temporally ordered in any way, the DFA must recognize these events for any ordering of the events. This means if a regular expression has some inherent ordering of simultaneous events, the DFA must accept all different permutations of such order. We refer to those permutations as linearizations. To build such a DFA, we augment the original DFA with alternative transitions to provide paths from one linearization to another. We then prove that the original DFA and the augmented DFA are equivalent. Only those DFAs whose underlying regular expressions have candidate simultaneous events in their relevant sets are augmented. The candidate simultaneous events can be statically detected by building and traversing a dependency graph described in Section 4.

The paper is organized as follows. Section 2 briefly explains an overview of the MaC framework. Section 3 introduces an extension MEDL-RE. Section 4 discusses the construction of the dependency graph. Section 5 presents and proves our augmented DFA algorithm. Section 6 presents related work. Lastly, section 7 concludes the paper.

## 2 MaC Overview

#### 2.1 MaC Architecture

The MaC system has been developed to ensure that a program runs correctly with respect to its formal requirement. Fig. 1 shows the MaC architecture. The system works as follows. A user specifies a requirement of a target program in a formal language. Given a target program and the requirement, the MaC system inserts a collection of probes or a *filter* into the target program. During run-time, the execution of the probed target program is monitored and checked by the MaC system. An *event recognizer* detects primitive events and conditions from state information received from the *filter*. The primitive events

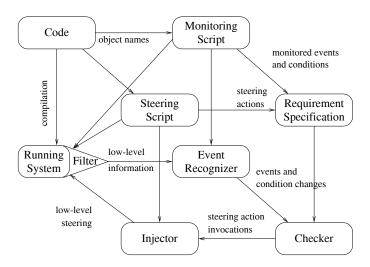


Fig. 1. Overview of the MaC architecture

are change of value (update(object)), entering method (startM(method)), and leaving method (endM(method)). The primitive conditions are boolean variables or boolean statements composed by primitive typed variables in the target program. These events and conditions are then sent to a run-time checker, which determines whether or not the current execution history satisfies the requirement specification. The execution history is captured from a sequence of events sent by the event recognizer. If the run-time checker detects any violation, it notifies the user and triggers an injector to take a steering action specified in a steering script and steer the target program back to a safe state.

#### 2.2 MaC Languages

The MaC system provides three languages. The requirement specification or the Meta-Event Definition Language (MEDL), based on an extension of a linear temporal logic (LTL) [13], allows us to express a large subset of safety properties of systems, including real-time properties. The monitoring script, expressed in the Primitive Event Definition Language (PEDL), is used to define what information is sent from a filter to the MaC system and how it is transformed into events and conditions used in MEDL. The steering script written in the Steering Action Definition Language (SADL) is used to specify actions to be invoked when violations occur. See [9,10] for PEDL and SADL details.

#### 2.2.1 Events and Conditions

Events occur instantaneously during the system execution, whereas conditions represent information that hold for a duration of time. For example, an event denoting the call to method init occurs at the instant the control is passed to the method, while a condition (angle < 30) holds as long as the value of the variable angle does not exceed 30. The syntax of events and conditions

is shown below.

```
E ::= e \mid \text{start}(C) \mid \text{end}(C) \mid E \&\& E \mid E \mid\mid E \mid E \text{ when } C
C ::= c \mid \text{defined}(C) \mid [E, E) \mid !C \mid C \&\& C \mid C \mid\mid C \mid C \Rightarrow C
```

The boolean connectives used in events and conditions are defined in the usual way. Events  $\mathtt{start}(c)$  and  $\mathtt{end}(c)$  are triggered when c becomes true and false, respectively. An event e when c is triggered when e is triggered and c is true. A condition  $\mathtt{defined}(c)$  is true when c is defined. A condition  $[e_1, e_2)$  is true between the occurrence of events  $e_1$  and  $e_2$  where  $e_1$  is included but  $e_2$  is not. For formal semantics of events and conditions, see [9,11].

# 2.2.2 Meta Event Definition Language (MEDL)

MEDL includes events and conditions imported from PEDL, definitions of composite events and conditions, safety properties, auxiliary variables, and auxiliary functions. A safety property can be expressed as a condition or as an event called an alarm. A safety property condition must always be true during the execution whereas an alarm must never be raised. Auxiliary variables can be used to define events and conditions. Auxiliary variables allow us, for example, to count the number of occurrences of an event. Auxiliary functions are time(e) and value(e), which return the time stamp and the value of an event e, respectively.

# 3 Syntax and Semantics of MEDL-RE

This section describes the extension of the MEDL language to include the specification of the ordering of events by expressing them as a regular expression and a frequency of events in a time interval.

#### 3.1 Syntax

Let  $\mathcal{R}$  be a set of regular expression names, e.g.,  $R, R_1, R_2$ , etc. Let s be statements in the MEDL-RE,  $s_{MEDL}$  be the existing MEDL statements,  $\bar{E}$  be a list of events, r be regular expressions, and C be the conditions described in Section 2.2.1. We define the syntax of the MEDL-RE extension as follows.

```
\begin{array}{lll} s & ::= s_{MEDL} & & \text{Existing MEDL statements} \\ & \mid \text{RE } R \mid \{\bar{E} \mid \} = < r > & \text{New MEDL-RE statement} \\ r & ::= E \mid r \cdot r \mid r + r \mid r * & \text{Regular expressions} \\ E & ::= e \mid \text{start}(C) \mid \text{end}(C) & \text{Existing events} \\ & \mid E \&\& E \mid E \mid E \mid E \text{ when } C \\ & \mid \text{startRE}(r) \mid \text{success}(r) \mid \text{fail}(r) & \text{Regular expression events} \\ f & ::= \text{time}(E) \mid \text{value}(E) \mid \text{occur}(E,C) & \text{Auxiliary functions} \end{array}
```

The MEDL-RE extension includes regular expressions, events  $\mathtt{startRE}(r)$ ,  $\mathtt{success}(r)$ ,  $\mathtt{fail}(r)$ , and an auxiliary function  $\mathtt{occur}(E,C)$ . The regular expressions, ranged over a set  $\Sigma$  of events, consists of events (E), concatenation

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$$\Sigma_{E} = \{E\}$$

$$\Sigma_{R_1+R_2} = \Sigma_{R_1} \cup \Sigma_{R_2}$$

$$\Sigma_{R_1 . R_2} = \Sigma_{R_1} \cup \Sigma_{R_2}$$

$$\Sigma_{R^*} = \Sigma_{R}$$

Fig. 2. A function  $\Sigma_R$ , which returns the relevant event set of R

 $(r \cdot r)$ , union (r+r), and Kleene star (r\*). The event  $\mathtt{startRE}(r)$  indicates that we start observing the regular expression. The event  $\mathtt{success}(r)$  indicates that we have found a sequence of events that specifies the regular expression, and the event  $\mathtt{fail}(r)$  indicates that we have started observing but failed to finish finding such a sequence. For an event E and a condition C, the auxiliary function  $\mathtt{occur}(E,C)$  returns a frequency or a number of occurrences of an event E during the time interval when C holds true.

#### 3.2 Semantics

Let  $\Sigma_R$  in Fig. 2 denote a set of events specified in a RE R and  $\Sigma_V$  denote a customized set of events specified as  $\bar{E}$  in the new MEDL-RE statement shown in the syntax section. Then, a relevant event set of R is  $\Sigma = \Sigma_R \cup \Sigma_V$ . We modify the model M defined in [9] as follows. A model M is a tuple  $(S, \tau, L_C, L_E, o)$ , where  $S = \{s_0, s_1, \ldots\}$  is a set of states,  $\tau$  is a mapping from S to the time domain,  $L_C$  is a total function from  $S \times C$  to  $\{true, false, \Lambda\}$  where C denotes a set of condition names and  $\Lambda$  denotes undefined, and  $L_E$  is a partial function from  $S \times E$  to a value domain where E denotes a set of event names. For all  $e_k$  where  $L_E(s_i, e_k)$  is defined, there is an order  $o(s_i, e_k)$  such that at time  $\tau(s_i)$ , an order for each occurrence  $e_k$  is distinct. The order o is a total and injective function that maps  $e_k$  and  $s_i$  to an ordered set of positive integers and  $o(s_{i-1}, e_k) < o(s_i, e_l)$  for all i and any k, l.

The semantics of MEDL-RE is defined using a derivative of a RE [5]. For any RE R and any alphabet a, a derivative of R with respect to a, denoted by  $D_a(R)$ , is the RE, where  $D_a(R) = \{x \in \Sigma^* | ax \in R\}$ . Fig. 3 and Fig. 4 show the semantics of a derivative of a RE and the function E(R) from  $\Sigma^*$  to boolean, which tests whether  $\epsilon \in R$ , respectively, where  $R_1, R_2$  are regular expressions,  $a, b \in \Sigma$  and  $a \neq b$ .

Besides the derivatives, we also need to define a function  $\operatorname{FIRST}(R)$  and a function  $\Phi_M^o(R)$ .  $\operatorname{FIRST}(R)$  returns a set containing all events that can appear as the first event in the RE R. Formally,  $\operatorname{FIRST}(R) = \{a \in \Sigma \cup \{\epsilon\} \mid ax \in R \text{ where } x \in \Sigma^*\}$ .  $\Phi_M^o(R)$  represents the remainder of the RE R at an order o after a sequence of derivatives. We define  $\Phi_M^o(R)$  as follows.

$$\begin{array}{ll} \Phi_M^{o(s_i,e)}(R) = R & \text{if } M, \tau(s_i) \models e \text{ where } e \in \texttt{FIRST}(R) \\ \Phi_M^{o(s_i,e)}(R) = D_e(\Phi_M^{o(s_i,e)-1}(R)) & \text{if } M, \tau(s_i) \models e \end{array}$$

We also define a language  $\mathcal{L}(R)$  of R as follows.

$$\mathcal{L}(\emptyset) = \emptyset$$
  $\mathcal{L}(\epsilon) = \{\epsilon\}$ 

$$\begin{array}{lll} D_a(a) & = \epsilon \\ D_a(b) & = \Lambda \\ D_a(\Lambda) & = \Lambda \\ D_a(\epsilon) & = \Lambda \\ D_a(R_1 + R_2) & = D_a(R_1) + D_a(R_2) \\ D_a(R*) & = (D_a(R)) \cdot R^* \\ D_a(R_1 \cdot R_2) & = (D_a(R_1)) \cdot R_2 & \text{if } E(R) = \text{False} \\ & & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\$$

Fig. 3. A derivative of a regular expression R with respect to  $a(D_a(R))$ 

$$\begin{array}{ll} E(a) & = \text{False} \\ E(\Lambda) & = \text{False} \\ E(\epsilon) & = \text{True} \\ E(R_1 + R_2) = E(R_1) \lor E(R_2) \\ E(R_1 \cdot R_2) & = E(R_1) \land E(R_2) \\ E(R^*) & = \text{True} \end{array}$$

Fig. 4. A function E(R), which tests whether  $\epsilon \in R$ 

$$\mathcal{L}(a) = \{a\} 
\mathcal{L}(R_1 + R_2) = \mathcal{L}(R_1) \cup \mathcal{L}(R_2) 
\mathcal{L}(R_1.R_2) = \{x_1.x_2 | x_1 \in \mathcal{L}(R_1) \land x_2 \in \mathcal{L}(R_2)\} 
\mathcal{L}(R^*) = (\mathcal{L}(R))^*$$

Using FIRST(R) and  $\Phi_M^o(R)$ , we define the startRE(R) event, success(R) event, and the fail(R) event.

$$\begin{array}{lll} \textit{M}, t &\models \texttt{startRE}(R) \text{ iff } \textit{M}, t \models e \text{ where } e \in \texttt{FIRST}(R) \\ \textit{M}, t &\models \texttt{success}(R) \text{ iff } \textit{M}, t \models e \text{ and } \epsilon \in \texttt{FIRST}(D_e(\Phi_M^{o(s_i,e)-1}(R))) \text{ where } t = \tau(s_i) \\ \textit{M}, t &\models \texttt{fail}(R) & \text{iff } \textit{M}, t \models e \text{ and } D_e(\Phi_M^{o(s_i,e)-1}(R)) = \Lambda \text{ where } t = \tau(s_i) \end{array}$$

The event startRE(R) is triggered when an occurring event is one of the events that can appear as the first event in the RE R. The event success(R) is triggered when an occurring event e causes the derivative of the remainder to contain empty string, and the event fail(R) is triggered when an occurring event e causes the derivative of the remainder to become undefined.

Next, we define a frequency function  $\operatorname{occur}(e, c)$ . A function  $\operatorname{occur}(e, c)$  returns the number of occurrences of an event e during the time interval that a condition c holds true.  $\operatorname{occur}(e, c)$  is defined as follows. Let  $f(s_i, e, c)$  denote a frequency of e in c at time  $\tau(s_i)$ . At time  $\tau(s_i)$ ,  $\operatorname{occur}(e, c) = f(s_i, e, c)$ .

$$f(s_i, e, c) = \Lambda \qquad \text{if } M, \tau(s_i) \not\models c$$

$$= 0 \qquad \text{if } M, \tau(s_i) \models start(c)$$

$$= f(s_{i-1}, e, c) \qquad \text{if } M, \tau(s_i) \models c \text{ and } M, \tau(s_i) \not\models e$$

$$= f(s_{i-1}, e, c) + 1 \text{ if } M, \tau(s_i) \models c \text{ and } M, \tau(s_i) \models e$$

#### 3.3 Examples

Two examples of requirements that need such timing dependencies are

(i) an event a must not occur three times in a row, and

(ii) the ordered events w,x,y,z can occur out of order for less than ten times when a condition c holds true.

Events are a, b, w, x, y, and z where events a and b are not related to events w, x, y, and z, and vice versa. To specify those requirements in the existing MEDL, we need a few auxiliary variables to keep track of when and how many times events a, w, x, y, and z occur. The fragment of MEDL below shows how we specify the above requirements in the original MEDL.

The aCount variable keeps track of how many times an event a occurs. Whenever a occurs, we increment it and whenever b occurs, we reset it. The alarm a3 alerts users when aCount becomes 3. The event wxyz is triggered only when c holds true and when y or z occurs between w and x or when z occurs between x and y. The wxyzCount variable stores the number of times the event wxyz has occured. The property wxyz10 alerts users when wxyzCount is greater than 10.

Consider when we need to specify the ordering of more than four events, the event such as wxyz can get very complicated because too many cases need to be considered. This shows that writing requirements using the existing MEDL is error-prone and difficult to understand. By adding regular expressions to the language, an order of events can be expressed more intuitively. The below fragment of the specification shows how to specify the example requirements in the extended MEDL-RE.

```
RE a3RE {b} = <a.a.a>;
RE wxyzRE {} = <w.x.y.z>;
alarm a3 = success(a3RE);
property wxyz10 = occur(fail(wxyzRE),c) < 10;</pre>
```

The regular expression a3RE and wxyzRE denote the ordering of events a occurring three times in a row and the ordering of the event w followed by the events x, y, and z. The relevant set {b} in a3RE indicates that if an event b occurs between two events a, then this sequence would fail to match a3RE. However, the MEDL-RE would ignore an event x if it occurs between two events a and would not fail the sequence. The relevant set {} in wxyzRE works similarly. The alarm a3 alerts users when we successfully match the RE a3RE while the property wxyz10 alerts users when the RE wxyzRE fails more than ten times. The requirements now are much simpler and easier to understand than the original MEDL.

# 4 Implementation

An important property of monitoring is its ability to monitor target systems as efficiently and quickly as possible using minimal system resources. Hence, using a deterministic finite automaton (DFA) to monitor a RE is preferred over a non-deterministic finite automaton (NFA). Several existing algorithms have been proposed to efficiently construct and minimize a DFA from a given RE. We have chosen the algorithm by Aho, Sethi and Ullman [2] because it generates a DFA directly without generating an NFA. The empirical result by Watson [15] also suggests that this DFA construction is efficient.

Simulation of a DFA involves the following steps. Identify a RE that has the current event in its relevant set. Then compare the current event with all the possible transitions from the current state of the DFA and take the appropriate one. If the current event initiates the simulation of a DFA, then it would trigger a startRE event. If the automaton moves to a final state, a success event is triggered under certain conditions as specified in Section 5. Similarly, the event causing the automaton to get stuck triggers a fail event. However, the problem in DFA simulation arises when there are simultaneous events generated by one primitive event or condition. Since there is no inherent order amongst these simultaneous events, the DFA simulation must evaluate multiple paths simultaneously for all possible permutations of these events. This is expensive in terms of resource utilization and may also be non-viable for certain applications especially in real time systems.

#### 4.1 Causes of Simultaneous Events

The MaC language consists of high-level or composite events and conditions, and low-level or primitive events and conditions described in Section 2.1 and 2.2. When primitive events or conditions are detected, the filter sends them to the event recognizer in the order as they occur. These primitive events and conditions can trigger composite events or change the value of conditions. While the primitive events and conditions are ordered, we cannot order the composite events triggered by one primitive event or conditions because all of the composite events occur at the same time.

Since REs can be defined on composite events that can occur individually and simultaneously, we need a way to recognize the events that can occur simultaneously. If two events can occur simultaneously, then their order is insignificant and their concatenation can be permuted without changing the meaning. To handle simultaneous events inexpensively, we propose the following steps. During the static phase, we determine if events used in the REs could occur simultaneously. If such events exist, we augment the original DFA with alternative transitions. At runtime, we try to take a step using the original transitions only if there exists a set of events from among the relevant set of this RE that have occured simultaneously.

```
e1 = update(A.x) when A.c1;
e2 = update(A.x) when A.c1 && A.c2;
e3 = update(A.y) when A.c2;
RE test {} = e1 . e2 . e3;
```

Fig. 5. Monitoring script

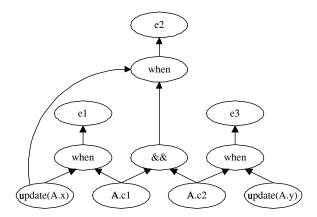


Fig. 6. Dependency graph

#### 4.2 Detecting Simultaneous Events

To detect possible simultaneous events statically, we construct a dependency graph from the syntax of events and conditions using connectives described in Section 2. The dependency graph G = (V, E) is a directed graph where the vertices V are connectives, events, and conditions, and the edges E represent dependency between them. Let  $e_1$ ,  $e_2$  be events or conditions. Then,  $(e_1, e_2) \in E$  if and only if  $e_1$  is used to compose  $e_2$ . For example, if  $e_3 = e_1||e_2$ , then  $(e_1, ||)$ ,  $(e_2, ||)$ , and  $(||, e_3)$  are in E. Fig. 5 shows an example monitoring script, and Fig. 6 shows the corresponding dependency graph. In the figures, there are two primitive events update (A.x), update (A.y) and two primitive conditions A.c1, A.c2. All of them are vertices at the bottom of Fig. 6. Since the event update (A.x) and the condition A.c1 are used to compose a connective when, edges (update (A.x), when) and (A.c1, when) are in E. Since the event e1 is composed of the connective when, an edge (when, e1) is also in E. The events e2, e3 are constructed similarly.

We denote that an event  $e_2$  depends on an event  $e_1$  if and only if there exists a path from an event  $e_1$  to an event  $e_2$  in G. We also denote that if there is a path to each of  $e_1$  and  $e_2$  from a common vertex in the graph, then they can occur simultaneously. Thus, a set of events that can be reached from the same primitive event in the dependency graph can all occur simultaneously. For example, in Fig. 6, the events  $e_1$ ,  $e_2$  depend on the same primitive event  $e_1$ , and therefore, they can occur simultaneously. To detect dependency, we traverse the graph in a bottom-up fashion using BFS.

However, the analysis of the dependency graph can yield a false positive result. This means that simultaneous events obtained from the graph might never occur in the actual system. For example, consider the events in Fig. 6. Based on the dependency graph, the events e1 and e2 can occur simultaneously because both depend on the event update(A.x). Suppose that the conditions A.c1 and A.c2 are boolean variables in the target program and they are never true at the same time. Then, events e1 and e2 never occur simultaneously.

There are two special cases. The first case is when an event is composed using an and connective  $(e_1 \&\& e_1)$ . The event with the && connective is triggered if and only if both of its two arguments  $e_1$  and  $e_2$  occur simultaneously. Therefore, if there exists an && connective along the path, the two arguments of the && connective must depend on one common vertex. Otherwise, the event with the && connective will never be triggered and therefore, can be eliminated from possible simultaneous events. The second case is when an event is composed using a when connective (e when c), which consists of an event side e and a condition side c. The event with the when connective is triggered if and only if the event e occurs when the condition c is true. Assume the two events we are detecting are  $e_1$  and  $e_2$ , and  $e_1$  is composed using a when connective. If the common vertex of  $e_1$  and  $e_2$  lies on the event side of the when connective in  $e_1$ , then  $e_1$  and  $e_2$  can occur simultaneously. If it lies only on the condition side, then  $e_1$  and  $e_2$  cannot occur simultaneously. For example, the events e2 and e3 in Fig. 6 cannot occur simultaneously. This is the case because update(A.x) and update(A.y) cannot occur simultaneously based on our assumption that all primitive events are always ordered. However, the events e1 and e2 can occur simultaneously if both A.c1 and A.c2 are true at the same time.

A set of possible simultaneous events obtained by the above process can be used to identify REs that would require augmentation. Only those regular expressions that have a subset of simultaneous events in their relevant sets need to be augmented.

# 5 Algorithm for Augmenting and Simulating DFA with Simultaneous Events

Based on the information of simultaneous events provided in Section 4, we incorporate additional information into the DFA statically. We claim that this additional information can assist the simultaneous simulation of multiple paths in the DFA efficiently. The following algorithm is proposed to incorporate this additional information into the DFA.

Let the DFA be a minimal DFA generated using the algorithm for DFA generation and minimization as described in Section 4. We call this DFA as  $DFA_{org}$  and the augmented DFA generated by the algorithm as  $DFA_{aug}$ . We construct  $DFA_{aug}$  under the following premises.

(i) The set of simultaneous events generated by the event recognizer does

not constitute multiple occurrences of the same event. This assumption is valid because the MaC system does not record the number of occurrences of an event at any given time instant. It only records the presence or absence of the event occurrence.

(ii)  $DFA_{org}$  consists of a single accepting state. Since we detect only the shortest sequence accepted by  $DFA_{org}$ , we can remove all outgoing transitions from all the accepting states. The minimization procedure would then merge those states into one.

Let  $DFA_{org} = (Q, \Sigma, T, q_0, q_f)$  where Q is a set of finite states,  $\Sigma$  is a relevant event set of R as described in Section 3.2,  $T: Q \times \Sigma \to Q$  is a partial transition function,  $q_0 \in Q$  is the start state, and  $q_f \in Q$  is the single accepting state. Given  $DFA_{org}$ , the algorithm constructs  $DFA_{aug} = (Q, \Sigma, T, q_0, q_f, D, U, ES, T_{alt})$  where  $Q, \Sigma, T, q_0$ , and  $q_f$  are obtained from  $DFA_{org}$ ,  $D: T \to \mathbb{N} \cup \{0\}$  is the distance annotation of a transition indicating its longest distance from  $q_f$ ,  $U: T \to \mathbb{N} \cup \{0\}$  is the unique distance function that maps each transition of a state to a unique natural number or zero,  $ES: Q \to \mathcal{P}(\Sigma^*)$  is an event string annotation of a state, and  $T_{alt}: Q \times \Sigma \to Q$  is a partial alternative transition function. We denote  $T(q, e) = \Lambda$  iff  $(q, e) \notin dom(T)$  and denote  $\langle q, e, q' \rangle \in T$  iff T(q, e) = q'. D, U, ES, and  $T_{alt}$  are described in further details in the following sections. The algorithm for  $DFA_{aug}$  construction has two phases: Annotation of transitions and states, and generation of alternative transitions.

#### 5.1 Phase 1: Annotation of Transitions and States

During this phase, we annotate each transition  $t \in T$  of  $DFA_{org}$  with a numeric value denoted as D(t), such that the value

- (i) equals the distance of the transition t from the accepting state  $q_f$ . This distance is equal to the number of transitions to be traversed to reach  $q_f$  from the current state using this transition t.
- (ii) denotes the maximum of all such distances.

Transitions that constitute cycles or self loops are ignored during this phase. This distance generation phase of the algorithm is shown in Fig. 8. This annotation can be done by traversing the DFA using a BFS algorithm starting at  $q_f$ .

This distance information D(t) will assist in ordering of transitions for simulation at runtime. We will traverse one of the outgoing transitions from the current state based on the distance values associated with the transitions, i.e., lower distance value first. The rationale behind this choice is that transitions that are closer to the accepting state should be attempted before those that are farther away from it. In case of a conflict between transitions (i.e., two transitions having the same distance value), we pick transitions based on the ordering of corresponding events in the relevant set. In Fig. 8 the start

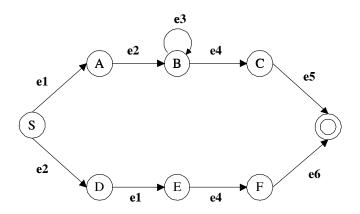


Fig. 7. Original minimized  $DFA_{org}$ 

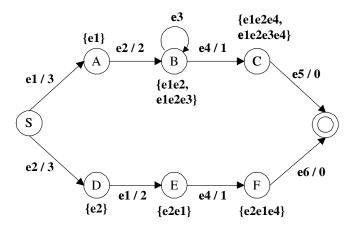


Fig. 8. Distance D and event string ES annotations of  $DFA_{aug}$ 

state S has two outgoing transitions both numbered 3. We will order these transitions based on the ordering of events e1 and e2 in the relevant set. Let this unique ordering of transitions t be called U(t).

During this phase, we also annotate each state of  $DFA_{aug}$  with additional information, indicating the set of possible sequences of input events that could be encountered to reach this state starting from the start state  $q_0$ . Each such input sequence is called the event string of that state. Please note that there can be multiple event strings for each state. This annotation step is shown in Fig. 8. The event string set for each state q, called ES(q), is equal to the concatenation of the set of event strings of all its predecessor states with the corresponding transitions. All states that have self loops on an event e have the event e appended to each of their event strings after the event strings are replicated. As shown in Fig. 8, the state e has a self loop on event e 1. Its set of event strings are then given by the event string e 1e1e2 for the normal transition and the event string e1e2e3 for the self loop. In case there is a cycle e1e2e2, then e1e2e3 for the event strings of the states that constitute the cycle after replication of the event strings. Please note that determination of event strings proceeds to the next state only after it is

```
1
     for each state q \in Q - \{q_0, q_f\}
2
         for each event string es \in ES(q)
3
            for each event e \in \Sigma such that T(q, e) = \Lambda
               for each state q_1 \in Q - \{q_0\} such that
4
               \max_{\forall e_i \in \Sigma, T(q_1, e_i) \neq \Lambda}(U(T(q_1, e_i))) \geq \min_{\forall e_i \in \Sigma, T(q, e_i) \neq \Lambda}(U(T(q, e_i)))
5
                  for each event string es_1 \in ES(q_1)
6
                     if (Hashed(es_1) == Hashed(es))
7
8
                        then Add \langle q, e, T(q_1, e) \rangle to T_{alt} (if T(q_1, e) \neq \Lambda)
                     else if (Hashed(es) == Hashed(Remove(e, es_1)))
9
10
                          then Add \langle q, e, q_1 \rangle to T_{alt}
```

Fig. 9. Algorithm to generate alternative transitions  $T_{alt}$ 

completely determined for the present state, i.e., all self loops for the current state must be resolved before proceeding to the next state. Cycles would be an exception here because we cannot identify cycles until we proceed ahead and return back to the state. Even in this case, the algorithm will proceed to states following the cycle only after all the event strings for all the states in the cycle have been determined.

#### 5.2 Phase 2: Generation of Alternative Transitions.

Before we proceed, we need to define the terms linearization and equivalence of linearization.

Linearization Linearization is a total order of a set of events. If this set consists of events that have occured simultaneously then the ordering of such simultaneous events in the set is one linearization of this set. A different ordering of these simultaneous events then constitutes a different linearization of this set.

Equivalence Two linearizations are equivalent if and only if they are different permutations of the same set of events.

During this phase, we add alternative transitions to appropriate states in  $DFA_{org}$ . The alternative transitions take the DFA from a state representing one linearization to another state representing an equivalent linearization for some set of simultaneous events. Let Remove(e, es) be an event string obtained by removing one occurrence of e from an event string es. Let Hashed(es) be a hash value associated with an event string es where the ordering of events in es is insignificant. This means two event strings have the same hash value if and only if they are equivalent linearizations. For example, Hashed(abc) = Hashed(cab) but  $Hashed(abc) \neq Hashed(abc)$ .

The algorithm shown in Fig. 9 compares the event strings of two states to see whether they are different linearizations of the same input sequence. If so, we say that the two event strings are equivalent linearizations of the input sequence, and we could add alternative transitions from one state to another without changing the meaning of the DFA. Since we always choose a transition t from a state q with the lowest U(t) first, the order of paths taken is unique. Thus, we need to add alternative transitions only from a state with lower U(t) to the states with higher U(t). This means that if q and  $q_1$  are states, then we add alternative transitions from q to  $q_1$  if and only if the maximum value of  $U(t_1)$  among all transitions  $t_1$  of  $q_1$  is greater than the minimum value of U(t) among all transitions t of t (lines 4–5). When adding the alternative transitions annotated with an event t from t to t to the event t to the existing transitions in t do not have a transition annotated with the event t (line 3). Thus, the resulting finite automata is still deterministic.

There are two different cases which the algorithm handles. Assume  $es \in ES(q)$  and  $es_1 \in ES(q_1)$  are two event strings such that q has lower U(t) than  $q_1$ . The first case is when some event strings of the two states q and  $q_1$  have different but equivalent linearizations. For example, es = abc and  $es_1 = bca$ . Here the algorithm adds an alternative transition on the current event e from the state q to the state represented by  $T(q_1, e)$  (lines 7–8). This transition is added for each event e such that there are no transitions currently present from q on any of these events. This case is shown in Fig. 10. Here, q=F,  $q_1=C$ , es=e2e1e4,  $es_1=e1e2e4$  and e=e5. The algorithm then adds a transition on e5 to T(C, e5) which is the accepting state  $q_f$ .

The second case is when  $Remove(e, es_1)$  and es represent equivalent linearizations for the two states q and  $q_1$ . For example, if es = ba,  $es_1 = aeb$ , and the current event is e, then the removal of e from  $es_1$  returns an event string which is equivalent to es. Here the algorithm adds an alternative transition on the current event e from q to  $q_1$  (lines 9–10). Again the transitions are added only for those events that currently do not have transitions from q. This case is shown in Fig. 11. Here, q=F,  $q_1=C$ , es=e2e1e4,  $es_1=e1e2e3e4$  and e=e3. The algorithm then adds a transition on e3 to e3.

After completion of this phase, we can delete the event string information from all the states except the ones with D(t) values of 1. We call these states the penultimate states. Their event strings will be used to verify the acceptance of the input sequence. This process will be explained in the next section.

### 5.3 DFA Simulation at Runtime

At runtime, we simulate  $DFA_{aug}$  without the alternative transitions as long as there is no occurrence of simultaneous events. On the first occurrence of simultaneous events, we activate all the alternative transitions. From the current state, we then take the path which is closest to the accepting state,

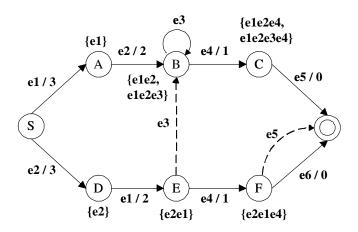


Fig. 10. Case 1 of the algorithm in Fig. 9

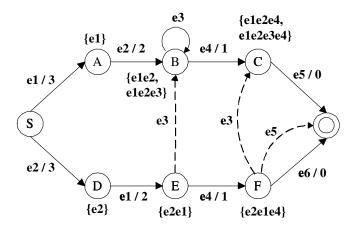


Fig. 11. Case 2 of the algorithm in Fig. 9

i.e., the path with the lowest U(t). Next, we simulate  $DFA_{aug}$  normally taking into consideration the alternative transitions and U(t).

Because events that occur simultaneously can also occur individually, some linearization that  $DFA_{aug}$  accepts might not be accepted by  $DFA_{org}$ . When  $DFA_{aug}$  reaches the accepting state, we verify by comparing the input sequence and the event strings in the penultimate states that have an original transition on the last event in the input sequence to the accepting state. Because  $DFA_{org}$  is minimal, only one such penultimate state exists. If they match,  $DFA_{aug}$  accepts that input sequence. Otherwise,  $DFA_{aug}$  rejects.

To be able to verify, we need to keep the history of all events seen from the start state. The history also keeps information about which events have occured simultaneously and which have occured individually. We do so by storing the input sequence as an array of simultaneous sets. To handle Kleene star, we have a flag to indicate whether or not each state has been visited. We then keep only events that take the simulation from one state to another unvisited state. This way, the resulting history string will resemble the way event strings are calculated and simplify our verification algorithm. Let

```
\begin{array}{ll} 1 & i=0, j=0 \\ 2 & \textbf{for each } e_i \in ES_{e_{last}} \\ 3 & \textbf{if } e_i \in h_j \\ 4 & \textbf{then } \operatorname{Remove } e_i \operatorname{ from } h_j \\ 5 & \textbf{else} \\ 6 & \textbf{then } \operatorname{Reject } \operatorname{this input } \operatorname{string} \\ 7 & \textbf{if } h_j \operatorname{ is empty} \\ 8 & \textbf{then } j=j+1 \\ 9 & \operatorname{Accept } \operatorname{this input } \operatorname{string } (\operatorname{if it } \operatorname{has not been } \operatorname{rejected}) \end{array}
```

Fig. 12. Verifying Algorithm

H = A history of the input sequence.

 $e_{last}$  = The last event in the input sequence.

 $ES_{e_{last}}$  = A set of event strings associated with the penultimate state, which has a transition in  $DFA_{org}$  on  $e_{last}$  to the accepting state.

 $e_i$  = A set of events in an event string where i is the position in the event string.

 $h_j$  = A set of simultaneous events  $h_j \in H$  where  $h_j$  occurs before  $h_{j+1}$ .

The verifying algorithm in Fig. 12 is straightforward. For each  $e_i \in ES_{e_{last}}$  (line 2), we check if  $e_i$  occurs in the current simultaneous set (line 3). If not, the input string is rejected (line 6). When the current set is completely parsed without being rejected, we move to the next set (lines 7–8). If we reach the end of  $ES_{e_{last}}$  without being rejected, we accept the input sequence (line 9). For example, let the input sequence recorded when the DFA accepted be  $\{e1e2, e4, e5\}$  where events e1 and e2 have occured simultaneously. Since the last event in the sequence, i.e.,  $e_{last} = e5$ , we have  $ES_{e_{last}} = \{e1e2e4, e1e2e3e4\}$ . Now since event string  $\{e1e2e4\}$  in  $ES_{e_{last}}$  matches with the recorded input sequence  $\{e1e2, e4\}$ ,  $DFA_{aug}$  will accept this input sequence.

## 5.4 Analysis of the Algorithm

This augmentation algorithm has the following running times during each phase. For phase 1, we run BFS twice, and thus the running time is O(s+t) where s=|Q| is the number of states, and t=|dom(T)| is the number of transitions. For phase 2, the running time is  $O(s^2 \cdot e \cdot es^2)$  where  $e=|\Sigma|$  is the number of events in the relevant set and  $es=\sum_{q\in Q}|ES(q)|$  is the number of possible event strings in the regular expression, (worst case analysis). This then constitutes the compile time penalties for augmenting the DFA. The

table in Fig. 13 compares the running time of this algorithm with the naive algorithm of dynamically simulating all possible paths. Since there can be  $O(2^n)$  such paths where n is the length of the input string, the dynamic simulation approach can impose an additional running time of  $O(2^n)$ . In our algorithm, O(n) added to the running time is for verification of the input sequence when  $DFA_{aug}$  accepts. The additional space is only  $O(n^2)$  because we add alternative transitions only from lower numbered to higher numbered paths, which is  $\sum_{i=1}^n i = O(n^2)$ . Therefore, our algorithm is clearly prefered over the naive algorithm.

Algorithm	Running Time	Space
Dynamic simulation of all paths	$O(2^n)$	0
Augmenting with alternative transitions	O(n)	$O(n^2)$

Fig. 13. Comparison between two algorithms

## 5.5 Equivalence of $DFA_{org}$ and $DFA_{aug}$

In this section, we provide a proof that  $DFA_{org}$  and  $DFA_{aug}$  are equivalent.

**Theorem 5.1**  $DFA_{aug}$  accepts if and only if  $DFA_{org}$  accepts.

#### Proof.

## Part 1: If $DFA_{org}$ accepts, $DFA_{aug}$ accepts.

This statement can be restated as if  $DFA_{org}$  accepts some linearization of the input sequence, then  $DFA_{aug}$  accepts all of its equivalent linearizations. We define a path and a set of paths as follows. A path in  $DFA_{org}$  is any path from the start state  $q_0$  to the accepting state  $q_f$ . Every such path is uniquely ordered by the algorithm. A set of paths is an enumeration of all possible paths from  $q_0$  to  $q_f$  of  $DFA_{org}$ . By the description of the algorithm, every alternative transition is a transition from one path to another path in the set, and the transitions are always taken from lower U(t) to higher U(t).

Let a be the event currently being considered. Let  $q_1$  and  $q_2$  be the two states being considered. Let  $es_1$  and  $es_2$  be the two event strings associated with  $q_1$  and  $q_2$ , respectively. Assume the current state is  $q_1$ , and therefore,  $es_1$  is one linearization of input seen so far. The algorithm has the following alternatives.

(i) The event string  $es_2$  is a different equivalent linearization of input seen so far, i.e.,  $(Hashed(es_1) = Hashed(es_2))$ . Then, the algorithm adds an alternative transition from  $q_1$  to a state  $q_3$  on a such that a transition from  $q_2$  to  $q_3$  on a exists, i.e.,  $T(q_2, a) = q_3$ . Thus, this alternative transition takes a step from one linearization to the other.  $(Hashed(concat(es_1, a)) = Hashed(concat(es_2, a)))$ .

(ii) The removal of a from  $es_2$  is a different equivalent linearization of the input seen so far.  $(Hashed(es_1) = Hashed(Remove(a, es_2)))$ . Then, this algorithm adds an alternative transition from  $q_1$  to  $q_2$  on a. Thus again this alternative transition takes a step from one linearization to the other.  $(Hashed(concat(es_1, a)) = Hashed(es_2))$ .

Now we prove that such alternative transitions exist for every pair of equivalent linearizations. Let states  $q_1$ ,  $q_2$ ,  $q_3$ ,  $q_4$  belong to paths  $p_1$ ,  $p_2$ ,  $p_3$ ,  $p_4$ , respectively, and let the ordering of these paths be  $p_1 < p_2 < p_3 < p_4$ . Then the ordering of states is  $q_1 < q_2 < q_3 < q_4$ . If one of the event strings associated with each of these states are different equivalent linearizations and the ordering is as given above, then it can be seen that the algorithm repeats itself for each pair of states  $(q_1, q_2), (q_1, q_3), (q_1, q_4), (q_2, q_3), (q_2, q_4)$  and  $(q_3, q_4)$  in that order. Thus, the algorithm provides a path to go from the current linearization of the input to any other different equivalent linearization governed by the unique ordering of linearizations (or paths).

Since at runtime we will simulate  $DFA_{aug}$  based on the ordering of the paths; from a given path of U(t)=k, we always have a path using the alternative transitions to all paths with U(t) greater than k and having some equivalent linearization. Thus,  $DFA_{aug}$  will eventually reach the path, which represents the linearization for which  $DFA_{org}$  accepts. Once it reaches this path, the process of picking original transitions first ensures that we will never leave this path.

# Part 2: If $DFA_{aug}$ accepts, $DFA_{org}$ accepts.

We prove this case by contradiction. Assume that  $DFA_{aug}$  accepts and  $DFA_{org}$  does not for some input sequence. By the definition of the algorithm,  $DFA_{aug}$  accepts if the following conditions exist.

- (i) Simulation reaches the accepting state of  $DFA_{aug}$ .
- (ii) The input sequence on which  $DFA_{aug}$  has accepted is equivalent to the concatenation of an event  $e_{last}$  and some event string associated with the penultimate state which has a transition on an event  $e_{last}$  to the accepting state, where  $e_{last}$  is the last event that occurs in the input sequence.

From the definition of event strings at a state, we can claim that the event string stored at the penultimate state is equal to a string parsed by  $DFA_{org}$  to reach this state. This claim is valid because while determining event strings, we use only the original transitions present in  $DFA_{org}$ . The transition taken from the penultimate state to reach the accepting state is also an original transition. Therefore, the event string concatenated by the last transition is accepted by  $DFA_{org}$ , contradicting the assumption above.

## 6 Related Work

There are a few existing works that incorporate regular expressions into logic. The ForSpec Temporal Logic (FTL) [3], Intel's new formal specification language, extends a linear temporal logic [13] with the ability to specify all  $\omega$ regular properties. FTL allows a user to define temporal connectives over time windows, regular sequences of Boolean events, and then relate such events via special connectives. Sugar [4] adds an extensive set of operators including regular expressions as a syntactic sugar to CTL [7]. Property Specification Language [1] is a specification language for hardware modeling and verification, which supports LTL [13] and extended regular expressions. Monitoring oriented Programming (MoP) [6] provides a monitoring architecture based on LTL [13] and extended regular expressions. Temporal Rover [8] is an architecture that helps a system do monitoring. Its specification language uses LTL [13] and MTL [12] with regular expressions and Time-Series. Time-Series observes temporal properties over time and is used for properties like stability, monotonicity, temporal average, sum, and max/min value. Comparing to the MEDL-RE based on LTL and regular expressions, FTL [3], MoP [6], and Temporal Rover [8] provide similar languages based on LTL and/or MTL and regular expressions while Sugar [4]'s language is based on CTL and similar regular expressions. However, none of them seem to have the issue of simultaneous events and therefore no algorithm for checking dependency and augmenting a DFA has been proposed.

## 7 Conclusion

We have presented an algorithm to simulate an augmented DFA for the extension MEDL-RE. The MEDL-RE incorporates regular expressions, which provide a more intuitive and less error-prone language to express complex dependencies between sequence of events, timing constraints, and a frequency of events during a time interval. The events associated with a regular expression offers the ability to detect the instance when the regular expression starts and the instance when we succeed or fail to find the regular expression. The DFA is augmented by adding alternative transitions after we statically detect the possibility of simultaneous events. We also prove that the augmented DFA is equivalent to the original DFA.

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