MONAA: a Tool for Timed Pattern Matching with Automata-Based Acceleration

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I. MONITORING

Monitoring over a real-time specification is an actively studied topic with a lot of industrial applications, such as monitoring of simulation traces of a Simulink model, and a HILS (hardware-in-the-loop simulation) with a system prototype. Given a log (a timed word or a signal) and a specification (a timed automaton (TA) [1], a timed regular expression (TRE) [2], or a formula in metric temporal logic [7]), a monitor finds all the segments of the log that satisfy the given specification.

A monitoring procedure has the online property if it starts the procedure before the entire log is given. This property is essential in monitoring a system that is currently running. The efficiency of a monitoring procedure is also important because recent trends such as autonomous driving have drastically increased the size of logs and the number of properties to monitor. Also, when the procedure is executed on a cloud server, an efficient monitor can reduce the pay-as-you-go cost.

One possible formalization of such monitoring problems is given by timed pattern matching, and both offline and online algorithms are proposed. See [9] and [10]: their theoretical results have led to their tool Montre [8]. Besides this series of works, the current authors have investigated efficient algorithms for timed pattern matching with automata-based acceleration: see [11] and [12]. The acceleration in our algorithms is based on the idea of skipping that comes originally from string matching (e.g., the KMP algorithm [6] and the BM algorithm [3]). The optimization there is by pre-computing a skip value table, and skipping unnecessary matching trials accordingly.

II. TIMED PATTERN MATCHING

We take an “event-based” formalization of timed pattern matching in [11, 12], unlike a “state-based” one in [9, 10]. To represent a log of a real-time system, we employ a timed word, which is a sequence of characters each of which is equipped with a real-valued timestamp.

Definition 1 (timed word). For an alphabet Σ, a timed word over Σ is a sequence $w = (a_1, \tau_1), (a_2, \tau_2), \ldots, (a_n, \tau_n) \in (\Sigma \times \mathbb{R}_{\geq 0})^*$ satisfying $\tau_i \leq \tau_{i+1}$ for any $i \in [1, n-1]$.

We let $w|_{(t, t')} \equiv (w(t, \tau(t)))_{t \in [t, t']}$ denote the restriction of $w$ to an interval $(t, t')$. See [12] for details.

To represent a real-time specification, we employ a TA, which is an NFA equipped with timing constraints. Since a TRE [2] can be translated to a TA [2], we can also use a TRE as a specification. The set of timed words accepted by a TA $A$ is denoted by $L(A)$.

Finally, our problem is formalized as follows.

Definition 2 (timed pattern matching). For a TA $A$ and a timed word $w$, the timed pattern matching problem asks for the set of matching intervals $\{(t, t') \mid w|_{(t, t')} \in L(A)\}$.

III. MONAA—A MONITORING TOOL ACCELERATED BY AUTOMATA

We present a tool MONAA for timed pattern matching. In MONAA, our timed FJS algorithm [12] is implemented. It has the online property and enjoys the constant speedup by skipping, typically twice or three times faster than without skipping. MONAA has two interfaces, the command-line interface MONAA and the C++ API libmonaa.

Algorithm Description: At the beginning of the timed FJS algorithm, it pre-constructs a skip value table. The table shows the number of matching trials to be skipped, depending on the observations obtained in the matching trial so far. The original FJS algorithm is for string matching [4]; there a skip value table is constructed comparing strings, exploiting finiteness of the pattern string. In our timed FJS algorithm—where a pattern is an infinite set $L(A)$ of words rather than a single string—defining a finite skip value table itself is a challenge. We use discrete states of TA for overapproximation, and construct a skip value table by checking emptiness of the intersection of the original TA $A$ and its variant where the initial state is shifted. In this process we crucially rely on TA constructions such as zones.

During the actual search, the timed FJS algorithm skips unnecessary matching trials using the pre-constructed skip value table. We remark that the runtime overhead of skipping is only by memory access and thus small.

The Command-Line Interface: In the command-line interface, MONAA reads a specification in either: a TA given in a file; or a TRE given as a command line argument. Reading a timed word from the standard input, MONAA writes the result of the timed pattern matching procedure to the
standard output. Since MONAA reads the timed word lazily, it can process a partial log provided by a system that is currently running. It can also notify a user of detection of matching behaviors before the whole matching is complete.

The C++ API: We also provide a C++ API called libmonaa. Because of the modularity, this API allows a user to write a program which performs the timed pattern matching procedure as part of the program. For example, one can implement a controller monitored in parallel, and the monitor changes the control mode when an unsafe behavior is detected.

In addition to the modularity, it also turns out that our C++ API is beneficial for performance. By hard-coding a TA in C++ code, we benefit from compiler optimization, and monitoring becomes faster.

IV. A Performance Comparison with Montre

We compare the performance of MONAA with that of the existing tool Montre [8], by monitoring real-time behaviors of a Simulink model from an automotive domain. The input timed words are generated from an automatic transmission model [5]. The input specification is the following TA (modulo minor rewriting for readability).

\[ ((g_1 \lor g_2 \lor g_3 \lor g_4 [\omega \geq 2500]) \lor (g_1 \lor g_2 \lor g_3 [\omega \geq 2500]) g_4) \]
\[ \lor (g_1 \lor g_2 [\omega \geq 2500] g_3 \lor g_4 [\omega \geq 2500]) g_1 \]
\[ \lor (g_1 \lor g_2 \lor g_3 [\omega \leq 2500]) (0,10) (g_3 \lor g_4 [\omega < 2500] [\omega \geq 2500])^+ (1,1000) \]

This means that the gear changes from the first (g_1) to the forth (g_4) and the engine rotation becomes high (\(\omega \geq 2500\)) within 10 seconds, and in the next second, the gear keeps being the third (g_3) or the forth (g_4) but the velocity does not get high (\(v \geq 100\)).

We compared MONAA giving either a TRE or a TA, and a libmonaa-based timed pattern matching program (in which a TA is hard-coded), with Montre’s online and offline modes. Our programs are compiled by GCC 7.1.0 with optimization flag -O3 and the experiments are conducted on an Amazon EC2 c4.large instance (January 2018, 2 vCPUs and 3.75 GiB RAM) that runs Ubuntu 16.04.2 LTS (64 bit).

The results of our experiments are in Table I and Table II. Table I shows that libmonaa-based monitor performs the fastest and the online mode of Montre performs the slowest. We remark that MONAA constantly takes about 7 seconds extra when a TRE is given. This is because of the translation from a TRE to a TA, which does not affect the remaining procedure. The execution time of MONAA grows only linearly with respect to the length of the input timed word, a characteristic desired for monitoring algorithms.

Table II shows that the memory usage of MONAA is independent of the length of the timed word, while that of Montre depends.

| Table I |
| --- |
| **EXECUTION TIME (sec.)** |
| Length of timed word | MONAA (TRE) | MONAA (TA) | MONAA (TA) in hard coded | Montre (online) | Montre (offline) |
| 708 | 7.05 | 0.80 | 0.50 | 0.35 | 0.07 |
| 218,447 | 7.55 | 1.27 | 0.31 | 0.35 | 0.08 |
| 2,058,511 | 8.05 | 1.71 | 0.42 | 0.45 | 0.13 |
| 655,237 | 8.54 | 2.21 | 0.53 | 115.84 | 4.69 |
| 870,967 | 9.16 | 2.69 | 0.64 | 153.71 | 6.21 |
| 1,287,411 | 9.53 | 3.14 | 0.75 | 189.55 | 7.75 |
| 1,304,404 | 10.05 | 3.60 | 0.85 | 216.92 | 9.33 |
| 1,527,612 | 10.51 | 4.10 | 0.95 | 269.77 | 10.48 |
| 1,739,525 | 11.05 | 4.56 | 1.07 | 289.63 | 12.39 |

| Table II |
| --- |
| **MEMORY USAGE (KBYTES)** |
| Length of timed word | MONAA (TRE) | MONAA (TA) | MONAA (TA) in hard coded | Montre (offline) |
| 708 | 16,468 | 10,808 | 7,588 | 27,456 |
| 218,447 | 16,312 | 10,808 | 7,641 | 67,784 |
| 2,058,511 | 16,344 | 10,692 | 7,390 | 87,689 |
| 655,237 | 16,486 | 10,840 | 7,298 | 99,540 |
| 870,967 | 16,547 | 10,980 | 7,432 | 109,070 |
| 1,287,411 | 16,700 | 10,768 | 7,292 | 147,048 |
| 1,304,404 | 16,840 | 10,696 | 7,440 | 153,992 |
| 1,527,612 | 16,846 | 10,690 | 7,440 | 169,600 |

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