

## A PROOF

### A.1 Proof of Proposition 1

Referring to Definition 1, a pattern  $p_0$  is declared over a set of events, say  $\mathcal{E}_0$ . It applies to the (at least two) sub-patterns  $p_1, \dots, p_k$  embedded in SEQ or AND of  $p_0$ , denoted by  $\mathcal{E}_1, \dots, \mathcal{E}_k$ . Referring to the sequence/concurrent relationships among  $p_1, \dots, p_k$ , we have  $\mathcal{E}_i \cap \mathcal{E}_j = \emptyset, i, j \in 1, \dots, k$ , i.e., the sub-patterns  $p_1, \dots, p_k$  do not overlap on events. It follows that the number of sub-patterns embedded in  $p_0$  is not greater than  $|\mathcal{E}_0|$ . As presented in Definition 2, the [ATLEAST  $a$ ] [WITHIN  $b$ ] conditions in each pattern can be checked in  $O(k)$  time. The time cost of determination of whether  $t \models p_0$  is thus  $O(|\mathcal{E}_0|^2)$ , where  $|\mathcal{E}_0|$  is the number of events declared in  $p_0$ .

### A.2 Proof of Theorem 2

The problem is clearly in NP. For any trace  $t$ , whether it matches each pattern can be recursively checked, referring to Proposition 1.

To show the hardness of the pattern consistency problem, we present a reduction from the 3SAT problem, which is one of Karp's 21 NP-complete problems [24]. Consider a CNF formula  $C = c_1 \wedge \dots \wedge c_m$ , where each clause  $c_i = l_{i1} \vee l_{i2} \vee l_{i3}$  have three literals,  $i = 1, \dots, m$ , and each literal  $l_{ij}$  is either  $x$  or  $\neg x$  for some variable  $x \in X, n = |X|$ . The 3SAT problem is to determine whether there exists an assignment to make all the clauses satisfied.

The transformation first introduces a set of events

$$\mathcal{E} = \{C_0, C_1, \dots, C_m, X_1, \dots, X_n, \neg X_1, \dots, \neg X_n\},$$

where each  $C_i$  corresponds to a clause  $c_i, i = 1, \dots, m$ , and  $X_j, \neg X_j$  represent  $x_j, \neg x_j$ , respectively,  $j = 1, \dots, n$ . A set  $\mathcal{P}$  of event patterns over  $\mathcal{E}$  are constructed as follows:

- (1) For each  $x_j, \neg x_j$ , we introduce an event pattern

$$p_j : \text{SEQ}(C_0, \text{AND}(X_j, \neg X_j)) \text{ ATLEAST } 1 \text{ WITHIN } 1) \text{ ATLEAST } 3 \text{ WITHIN } 3.$$

- (2) For each clause  $c_i = l_{i1} \vee l_{i2} \vee l_{i3}$ , we add an event pattern

$$p_{n+i} : \text{SEQ}(C_i, \text{AND}(X_{i1}, X_{i2}, X_{i3})) \text{ ATLEAST } 2 \text{ WITHIN } 2.$$

- (3) For each  $C_i$ , we consider

$$\text{SEQ}(C_0, C_i) \text{ WITHIN } 1.$$

Event pattern  $p_j$  requires that the timestamp distance between  $X_j$  and  $\neg X_j$  is exactly 1, while  $p_{n+i}$  requires that at least one of  $X_{i1}, X_{i2}, X_{i3}$  should have timestamp distance 2 compared to  $C_i$ . (Figure 14 illustrates the interval conditions in the complex temporal networks corresponding to the aforesaid patterns, see Definition 5 for details.)

We show that the CNF formula in transformation is satisfiable if and only if there exists a trace  $t$  of events such that  $t[C_0] = s, t[C_i] = s + 1, t[X_{ik}] = s + 3$  for some  $X_{ik}, k = 1, 2, 3, i = 1, \dots, n$ ,  $s$  is any timestamp in the domain, i.e.,  $t \models \mathcal{P}$  the event patterns are consistent.

First, if the CNF formula in transformation is satisfiable, we set the corresponding timestamp assignment  $t[C_i] = s + 1$  and  $t[C_0] = s$ , where  $s$  is any timestamp in the domain. For each clause  $c_i = l_{i1} \vee l_{i2} \vee l_{i3}$ , there must exist some  $l_{ik} = \text{true}, k = 1, 2, 3$ , while others equal 0. For each  $l_{ik} = \text{true}$ , we assign  $t[X_{ik}] = s + 3$ ; otherwise,

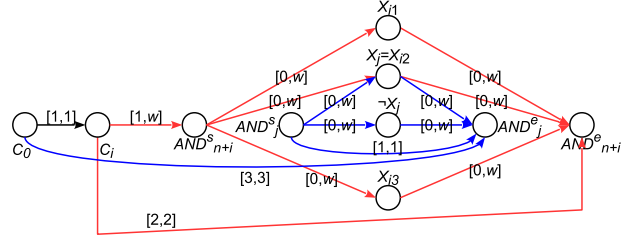


Figure 14: Reduction from 3SAT problem

$t[X_{ik}] = s + 2$  for  $l_{ik} = \text{false}$ . It is easy to see  $t \models p_{n+i}$  in  $\mathcal{P}$  in the transformation. Moreover, for each  $X_j = \text{true}$ , we have  $t[X_j] = s + 3$  and  $t[\neg X_j] = s + 2$ ; otherwise,  $t[X_j] = s + 2$  and  $t[\neg X_j] = s + 3$  for  $X_j = \text{false}$ . The CNF assignment guarantees that  $t \models p_j$  in  $\mathcal{P}$  in the transformation. That is, we have  $t \models \mathcal{P}$ .

Conversely, suppose that the event patterns are consistent, i.e., there exists a timestamp assignment  $t$  having  $t \models \mathcal{P}$ . Let  $t[C_0] = s$  the timestamp in the domain. It follows  $t[C_i] = s + 1, i = 1, \dots, m$ . Pattern  $p_j$  in  $\mathcal{P}$  in the transformation ensures that only one of  $X_j, \neg X_j$  will have  $\pi$  label  $s + 3$ , i.e., true in the CNF assignment; the other will have  $t$  label  $s + 2$  and false assignment in the CNF. Referring to  $t \models p_{n+i}$  in  $\mathcal{P}$  in the transformation, for each  $i = 1, \dots, n$ , there must exist some  $X_{ik}, k = 1, 2, 3$  such that  $t[X_{ik}] = s + 3$ . According to the aforesaid assignment, we have the corresponding literal  $l_{ik} = \text{true}$ . For others with  $t[X_{ik}] = s + 2$ , it has  $l_{ik} = \text{false}$ . That is, the CNF formula is satisfiable.

### A.3 Proof of Theorem 3

The problem is clearly in NP. Given a trace  $t'$  with modified timestamp, whether it matches each pattern can be recursively checked, referring to Definition 2, and the modification cost can be calculated in  $O(n)$  time by Formula 1.

To illustrate the hardness of the timestamp modification problem, we show a reduction from the SET COVER problem, which is one of Karp's 21 NP-complete problems [24].

Given a set of  $m$  elements  $\mathcal{U} = \{u_1, \dots, u_m\}$  and  $n$  sets  $\mathcal{S} = \{s_1, \dots, s_n\}$  such that  $s_i \subseteq \mathcal{U}$  and  $\cup_i s_i = \mathcal{U}$ . A set cover is a  $C \subseteq \mathcal{S}$  of sets whose union is still  $\mathcal{U}$ . The minimum set cover problem is to identify the smallest number of sets whose union still contains all elements in  $\mathcal{U}$ .

Consider a set of events

$$\mathcal{E} = \{S_1, \dots, S_i, \dots, S_n, S'_1, \dots, S'_n, U_1, \dots, U_j, \dots, U_m\}$$

where each  $S_i, S'_i$  represents a set  $s_i$ , and  $U_j$  corresponds to an element  $u_j$ . Let  $\mathcal{E}_j = \{S_i \in \mathcal{E} \mid u_j \in s_i\}$ , i.e., all the events  $S_i$  whose corresponding  $s_i$  contains  $u_j$ .

First, the event pattern set  $\mathcal{P}$  is constructed as follows.

- (1) For each  $U_j$ , we add into  $\mathcal{P}$  a pattern

$$p_j : \text{SEQ}(U_j, \text{AND}(S_{i1}, \dots, S_{ik})) \text{ ATLEAST } 2 \text{ WITHIN } 2,$$

where  $S_{i1}, \dots, S_{ik}$  correspond to all the sets  $s_{i1}, \dots, s_{ik} \subseteq \mathcal{S}$  containing element  $u_j$ .

- (2) For each  $U_j, S'_i$ , we introduce a pattern

$$\text{SEQ}(S'_i, U_j) \text{ WITHIN } 1.$$

(Figure 15 illustrates the interval conditions in the complex temporal networks corresponding to the aforesaid patterns, see Definition 5 for details.)

Next, we build a trace  $t$  of events where  $t[S'_i] = 0, t[U_j] = 1, t[S_i] = 2, i = 1, \dots, n, j = 1, \dots, m$ . The patterns require that there must exist a  $S_i \in \mathcal{E}_j$  such that  $t[S_i] - t[U_j] = 2$ , which is not observed in the given  $t$ .

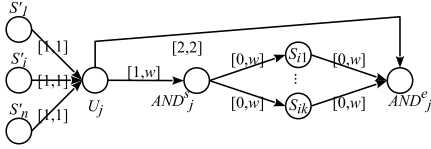


Figure 15: Reduction from SET COVER problem

To prove NP-hardness we show that there is a set cover  $C$  of size  $k$  if and only if there exists a trace  $t'$  matches the event patterns in the transformation and  $\Delta(t, t') = k$ .

First, if there exists a set cover  $C$  with size  $k$ , for each  $s_i \in C$ , we assign the corresponding  $t'[S_i] = 3$ . Referring to the repair cost in Formula 1, we have  $\delta(t', t) = k$ . Moreover, the set cover requires that for each  $u_j$ , there is at least one set  $s_i$  in  $C$  that contains  $u_j$ , i.e., having the corresponding  $t'[S_i] = 3$ . It leads to  $t' \models \mathcal{P}$ .

Suppose that there is a trace  $t'$  with  $t' \models \mathcal{P}$  and  $\Delta(t, t') = k$ . Referring to the transformation of each pattern  $p_j$  in  $\mathcal{P}$ , there must have some  $S_i \in \mathcal{E}_j$  with repaired  $t'[S_i] = 3$ . By collecting all the corresponding  $s_i$ , since all the elements  $u_j$  are covered, it forms a set cover  $C$  with size  $k$ .

#### A.4 Proof of Theorem 4

To show the hardness of approximation, we illustrate that the reduction in Theorem 3 from the SET COVER problem, which is known NP-hard to approximate to within any constant factor [10], is gap-preserving.

Let  $C^*$  denote a minimum set cover, and  $t^*$  be a modified trace with the minimum modification cost. It is sufficient to conclude

- if  $|C^*| \leq k$ , then  $\Delta(t, t^*) \leq k$ ,
- if  $|C^*| > \alpha k$ , then  $\Delta(t, t^*) > \alpha k$ ,

where  $\alpha > 1$ .

#### A.5 Proof of Proposition 5

The conclusion can be proved by induction referring to the recursive Definitions 2 and 5.

(1) For  $p = E$ , we have  $t[p^s] = t[p^e] = t[E]$  for any  $t$  in Definition 2, and  $\Phi_p = \emptyset, \Gamma_p = \emptyset$ , referring to item 1 in Definition 5.

(2) For  $p = \text{SEQ}(p_1, p_2, \dots, p_k) [\text{ATLEAST } a] [\text{WITHIN } b]$ , referring to item 2 in Definition 5, it has  $\Gamma_p \setminus (\Gamma_{p_1} \cup \dots \cup \Gamma_{p_k}) = \emptyset$ . First, suppose that  $t \models p$ . Referring to  $t[p_i^e] < t[p_{i+1}^e]$  in item 2 in Definition 2, we have  $t \models \phi(E_{p_i}^e, E_{p_{i+1}}^e); [1, w], i = 1, \dots, k-1$ , in item 2 in Definition 5. Moreover, the ATLEAST  $a$  WITHIN  $b$  condition is equivalent to the interval condition  $\phi(E_p^s, E_p^e); [a, b]$ . It leads to  $t \models (\Phi_p, \Gamma_p)$ . Conversely, we show  $t \models (\Phi_p, \Gamma_p)$  implies  $t \models p$  in a similar way.

(3) For  $p = \text{AND}(p_1, p_2, \dots, p_k) [\text{ATLEAST } a] [\text{WITHIN } b]$ , the binding condition  $\gamma(E_p^s, \{E_{p_1}^s, \dots, E_{p_k}^s\}); \text{min}$  and the interval conditions  $\phi(E_p^s, E_{p_i}^s); [0, w], i = 1, \dots, k$  in item 3 in Definition 5 ensure  $t[p^s] = \min(\pi(p_1^s), \pi(p_2^s), \dots, \pi(p_k^s))$  in item 3 in Definition 2. Similar equivalence on max binding condition can also be observed. Again, the ATLEAST  $a$  WITHIN  $b$  condition is equivalently captured by the interval condition  $\phi(E_p^s, E_p^e); [a, b]$ .

#### A.6 Proof of Corollary 6

Similar to the proof of Proposition 5, the conclusion can be proved by induction referring to the recursive Definitions 2 and 6. In particular, since there is no binding condition, the proof w.r.t. binding conditions in step (3) in the proof of Proposition 5 is not necessary.

#### A.7 Proof of Proposition 7

The binding conditions  $\Gamma$  appear only w.r.t. AND event patterns, together with a list of interval conditions, referring to item 3 in Definition 5.

First, suppose that there exists a  $\Phi_k \in \mathbf{\Phi}_\Gamma$  such that  $\Phi \cup \Phi_k$  is consistent, i.e., having some  $t \models \Phi \cup \Phi_k$ . For each binding condition (say  $\gamma(E_i, \mathcal{E}_i); \text{min}$ ) in  $\Gamma$ , we have some  $E_j \in \mathcal{E}_i$  with  $t[E_i] = t[E_j]$ , according to the full binding in Definition 7. Since  $t \models \Phi$ , we have  $t[E_i] = t[E_j] \leq t[E_k], \forall E_k \in \mathcal{E}_i$ , referring to the aforesaid interval conditions in item 3 in Definition 5. That is, it has  $t \models \Gamma$  as well.

Next, suppose that  $(\Phi, \Gamma)$  is consistent, having some  $t \models (\Phi, \Gamma)$ . For each binding condition (say  $\gamma(E_i, \mathcal{E}_i); \text{min}$ ) in  $\Gamma$ , we have some  $E_j \in \mathcal{E}_i$  with  $t[E_i] = t[E_j]$ , according to the binding condition Definition 4. That is, it satisfies  $t \models \phi(E_i, E_j); [0, 0]$  in  $\Phi_\gamma$ . Considering the aforesaid interval condition for each binding condition  $\gamma$  in  $\Gamma$ , we obtain a  $\Phi_k \in \mathbf{\Phi}_\Gamma$  having  $t \models \Phi_k$ .

#### A.8 Proof of Proposition 8

The conclusion is clear by considering all the pairwise shortest paths, over the distance graph with  $2 * (n-1)$  edges in two paths, respectively.

#### A.9 Proof of Proposition 9

Firstly, we prove  $\Phi_c \models \Phi$ . For any  $t \models \Phi_c$ , according to Formula 2, we have  $t[E_j] - t[E_i] = t[E_j] - t[E_{j-1}] + t[E_{j-1}] - \dots - t[E_i] \geq -d_{i+1,i} - d_{i+2,i+1} - \dots - d_{j,j-1} = -d_{ji} \geq a$  for any  $\phi(E_i, E_j); [a, b] \in \Phi, i < j$ . Similar derivation applies to  $t[E_j] - t[E_i] \leq d_{ij} \leq b$ . That is,  $t \models \Phi_c$  implies  $t \models \Phi$ .

Next, we prove  $\Phi \models \Phi_c$ . For any  $t \models \Phi$ , referring to the shortest path, we have  $-d_{i+1,i} \leq t[E_{i+1}] - t[E_i] \leq d_{i,i+1}, i = 1, \dots, n-1$ . Given the definition  $\phi(E_i, E_{i+1}); [-d_{i+1,i}, d_{i,i+1}]$  in  $\Phi_c$ , it follows  $t \models \Phi_c$ .

#### A.10 Proof of Proposition 10

To capture the bound of optimal timestamp modification, we first show that a trace timestamp explanation with optimal timestamp modification  $t^*$  always exists, such that for each  $E_i$ , it either has  $t^*(E_i) = t[E_i]$  or  $\exists E_j \in \mathcal{E}$  that  $t^*(E_j) = t[E_j], t^*(E_i) - t^*(E_j) = d_{ji}$  or  $-d_{ij}$ .

For each  $E_i$ , let  $\mathcal{E}_i = \{E_i\} \cup \{E_j \mid t^*(E_i) - t^*(E_j) = d_{ji} \text{ or } -d_{ij}\}$ . According to Definition 8 of chain temporal networks, we have  $\forall E_k \in \mathcal{E} \setminus \mathcal{E}_i, E_j \in \mathcal{E}_i, t^*(E_k) - t^*(E_j) \neq d_{jk}, t^*(E_k) - t^*(E_j) \neq -d_{kj}$ .

If there exists some  $E_j \in \mathcal{E}_i$  such that  $t^*(E_j) = t[E_j]$ , the conclusion is proved. Otherwise, i.e.,  $t^*(E_j) \neq t[E_j], \forall E_j \in \mathcal{E}_i$ , we can generate a new trace  $t'$  by slightly modifying a same small amount  $c$  for all the timestamp of events  $E_j \in \mathcal{E}$ . There are two cases:

(1) If  $|\{E_j \in \mathcal{E}_i \mid t^*(E_j) > t[E_j]\}| \neq |\{E_j \in \mathcal{E}_i \mid t^*(E_j) < t[E_j]\}|$ , the total modification cost of  $t'$  reduces, which is a contradiction to the optimal  $t^*$ .

(2) If  $|\{E_j \in \mathcal{E}_i \mid t^*(E_j) > t[E_j]\}| = |\{E_j \in \mathcal{E}_i \mid t^*(E_j) < t[E_j]\}|$ , where the total modification cost of  $t'$  is the same to  $t^*$ , there are two sub-cases:

(2a) We have  $t'[E_j] = t[E_j]$  for some  $E_j \in \mathcal{E}_i$ . The conclusion is prove.

(2b) We have  $t'[E_k] - t'[E_j] = d_{jk}$  or  $-d_{kj}$  for some  $E_k \in \mathcal{E} \setminus \mathcal{E}_i, E_j \in \mathcal{E}_i$ .  $E_k$  is then added into  $\mathcal{E}_i$  w.r.t.  $t'$  which has the same modification cost to  $t^*$ , i.e., optimal. We iteratively apply the aforesaid cases (1) and (2) for  $t'$ .

Finally, if all the events are added into  $\mathcal{E}_i$ , it belongs to either case (1) or (2a).

Referring to the aforesaid proved property of optimal solution, for any  $E_i \in \mathcal{E}$ , we have an optimal timestamp modification with  $t^*(E_i)$  no greater than  $t[E_j] + d_{ji}$  for all  $E_j \in \mathcal{E}$ . Similar lower bound applies to  $t[E_j] - d_{ij}$ . The conclusion is proved.

### A.11 Proof of Proposition 11

For the events  $\mathcal{E} = \{E_1, \dots, E_n\}$  in the pattern, we denote  $E_{\max} = \arg \max_{E \in \mathcal{E}} t[E]$  and  $E_{\min} = \arg \min_{E \in \mathcal{E}} t[E]$ . Referring to the single binding in Definition 9, we have

$$\Phi_1 = \{\phi(\text{AND}^s, E_{\min}):[0, 0], \phi(\text{AND}^e, E_{\max}):[0, 0]\}.$$

Let  $t^*$  be the optimal timestamp modification w.r.t. full binding, having  $E_{\max}^* = \arg \max_{E \in \mathcal{E}} t^*(E)$ , and  $E_{\min}^* = \arg \min_{E \in \mathcal{E}} t^*(E)$ . We show that  $t^*$  is also a timestamp modification w.r.t. the single binding, in all the possible cases.

(1) If  $t[E_{\min}] \leq t^*[E_{\min}^*]$ , it has  $t^*[E] = t^*[E_{\min}^*]$  for all  $E \in \mathcal{E} \setminus \{E_{\max}\}$  with  $t[E] \leq t^*[E_{\min}^*]$ . Any other possible timestamp modification on  $E$  greater than  $t^*[E_{\min}^*]$  must have modification cost no less than  $\pi^*$ . It indicates  $t^*[E_{\min}] = t^*[E_{\min}^*]$ .

(2) If  $t[E_{\min}] > t^*[E_{\min}^*]$ , it has  $t^*[E_{\min}] = t^*[E_{\min}^*]$ . Any other possible modification with a different event having the minimum timestamp  $t^*[E_{\min}^*]$  must have modification cost no less than  $\pi^*$ . It also indicates  $t^*[E_{\min}] = t^*[E_{\min}^*]$ .

(3) If  $t[E_{\max}] \geq t^*[E_{\max}^*]$ , it has  $t^*[E] = t^*[E_{\max}^*]$  for all  $E \in \mathcal{E} \setminus \{E_{\min}\}$  with  $t[E] \geq t^*[E_{\max}^*]$ . Any other possible timestamp modification on  $E$  less than  $t^*[E_{\max}^*]$  must have modification cost no less than  $\pi^*$ . It indicates  $t^*[E_{\max}] = t^*[E_{\max}^*]$ .

(4) If  $t[E_{\max}] < t^*[E_{\max}^*]$ , it has  $t^*[E_{\max}] = t^*[E_{\max}^*]$ . Any other possible modification with a different event having the maximum timestamp  $t^*[E_{\max}^*]$  must have modification cost no less than  $\pi^*$ . It also indicates  $t^*[E_{\max}] = t^*[E_{\max}^*]$ .

Since we have  $t^*[E_{\max}] = t^*[E_{\max}^*]$  and  $t^*[E_{\min}] = t^*[E_{\min}^*]$  in all the cases, it is sufficient to show that  $t^*$  is also a timestamp modification w.r.t. the single binding  $\Phi_1$ , and could be returned by Algorithm 2 with single binding.