The Algebra of Spatio-Temporal Intervals

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Abstract

The relations among temporal intervals can be used to model all time dependent objects. We propose a fast mechanism for temporal relation compositions. A temporal transitive closure table is derived, and an interval-based temporal relation algebraic system is constructed. Thus, we propagate the time constraints of arbitrary two objects across long distances n by linear time. We also give a complete discussion of different possible domains of interval relations. A set of algorithms is proposed to detect time conflicts and to derive reasonable interval relations. The algorithms are extended for time-based media in an arbitrary n-dimensional space.

1 Introduction

Communication networks and multimedia applications usually contains a number of resources to be presented sequentially or concurrently. Temporal interval relations represent the timing among resources. These resources need to be analyzed to ensure that there is no time conflict among resources. Moreover, many of these resources, occupy period of time and screen space. These data can be heavily time-dependent, such as audio and video in a motion picture, and can require time-ordered presentation. The spatio-temporal relations among resources need to be computed and represented.

The importance of knowledge underlying temporal interval relations was found in many disciplines. As pointed out in [1], researchers of artificial intelligence, linguistics, and information science use temporal intervals as a time model for knowledge analysis. For instance, in a robot planning program, the outside

world is constantly changed according to a robot's actions. The notion of "number three box is on the left of number two box" is true only within a temporal interval. The work discussed in [1] analyzes the relations among temporal intervals. However, the work [1] only states temporal interval relations. No spatial relation were discussed. We found that these relations can be generalized for spatial modeling.

We have surveyed many researches related to the spatio-temporal semantics of multimedia and distributed objects. However, no discussion of the conflict situation among relations were found. We found that, the use of spatio-temporal relations serves as a reasonable semantic tool for the underlying representation of objects in many multimedia applications. Composite objects can have arbitrary timing relationships. These might be specified to achieve some particular visual effect of sequence.

2 The Spatio-Temporal Relation Domains

According to the interval temporal relations introduced in [1], there are 13 relations ({e, <, >, m, mi, d, di, o, oi, s, si, f, fi}) between two temporal intervals. We describe the symbolic constraint propagation. The general idea is to use the existing information about the relations among time intervals or instants to derive the composition relations.

The composition may result in a multiple derivation. For example, if "X before Y" and "Y during Z", the composed relation for X and Z could be "before", "overlaps", "meets", "during", or "starts". If the composed relation could be any one of some relations, these derived relations are called reasonable relations in our discussion. A reasonable set is a set of reasonable relations according to our definition.

In some cases, relation compositions may result in a

conflict specification due to the user specification or involved events synchronously. For example, if specifications " X before Y "," Y before Z ", and " X after Z " are declared by the user, there exists a conflict between X and Z. When the specific relations are not found in derived reasonable set, the specification may cause conflicts.

We analyze the domain of interval temporal relations and use an directed graph to compute the relations of all possibilities.

Definition: An user edge denotes a relation between a pair of objects defined by the user. The relation may be reasonable or non-reasonable.

Definition: A derived edge holds a non-empty set of reasonable relations derived by our algorithm. The relation of the two objects connected by the derived edge can be any reasonable relation in the set.

For an arbitrary number of objects, some of the relations are specified by the user while others are derived. If there exists a cycle in the directed relation graph, a conflict derivation may occur. We suggest that the computation domain reveals four types, as discussed below.

- The complete relation domain (a complete graph): contains user edges and derived edges, with possible cycles and possible conflicts.
- The reasonable relation domain (a graph): contains user edges and derived edges, with possible cycles but no conflict.
- The reduced relation domain (a graph): contains only user edges, with possible cycles and possible conflicts
- The restricted relation domain (a tree): contains only user edges, without cycle.

The four domains are used in the analysis and computation of object relations. In section 4, we propose two algorithms computing the reasonable relation domain.

3 The Finite Temporal Relations Group

Based on Allen's work, transitivity table for the twelve temporal relations (omitting "=") showing the composition of interval temporal relations. Compositions of three or more relations are computed using algorithms based on set operations, such as set union and intersection. These set operations are expensive. We argue that, an extension of *Table13* (Allen's Table), named *Table29*, can be calculated. The com-

positions of three or more relations can be ob-tained directly from our table. Algorithm *Compute-Table29* calculates *Table29*, which consists of the compositions of 29 temporal relation sets. Based on the *table29*, we found many properties of spatio-temporal relations and proved the temporal relation composition is a algebraic *group*.

Firstly, we define some terminology. An interval has a name, which is an ASCII string. The term P(X) represents a power set of ob-jects of type X. The 13Rel is the domain of the 13 interval relations. Inverse relations are also defined. The 29Relset is a domain of relation sets. Each element in 29Relset contains one or more interval relations which represent the possible composition results between two intervals.

```
Name = = P(string)
13Rel == \{ <, >, d, di, o, oi, m, mi, s, si, f, fi, e \}
<^{-1} => \wedge d^{-1} = di \wedge o^{-1} = oi \wedge m^{-1} = mi
 \wedge s^{-1} = si \wedge f^{-1} = fi \wedge e^{-1} = e
29Relset \subset P(13Rel)
TemporalTuple = = Name \times 29RelSet \times Name
\forall tt: Temporal Tuple \bullet
        tt = (A, rs, B) \Leftrightarrow tt^{-1} = (B, rs^{-1}, A)
o: TemporalTuple × TemporalTuple →
       Temporal Tuple
\forall tt_1, tt_2, tt_3 : Temporal Tuple \bullet tt_1 = (A, rs_1, B) \land
 tt_2 = (B, rs_2, C) \wedge tt_3 = (A, rs_3, C) \bullet
tt_1 \bullet tt_2 = tt_3 \Leftrightarrow
    (A = C \land rs_2 = rs_1^{-1} \Rightarrow rs_3 = \{e\} \lor
    A = C \wedge rs_2 \neq rs_1^{-1} \Rightarrow rs_3 = \bot \vee
    A \neq C \Rightarrow rs_3 = Table 29 (rs_1, rs_2)
```

A temporal tuple contains two interval names and a relation set. The temporal tuple composition operator (i.e., o) checks whether the interval names A and C are equal. If so, and if the relation set of tt_1 and tt_2 are the inverse of each other, the composition results in an equality (i.e., $\{e\}$). On the other hand, if A and C are not equal, Table29 is used for looking up the composition result.

The following table gives a summary of the 29 relation sets which contain all possible composition results:

The 29 Relation Sets

ID	Relation	Sets	
1	{ < }		
2	{ > }		
3	{ d }		
4	{ di }		
5	{ o }		
6	{ oi }		
7	{ m }		
8	{ mi }		
9	{ s }		
10	{ si }		

```
11
            { f }
            { fi }
12
            { e }*
13
14
            { o, di, fi }
            { oi, d, f }
15
            { o, d,s }
16
            { oi, di, si }
17
18
            { <, o, m }
19
            { >, oi, mi }
20
            { f, fi, e }*
            { s, si, e }*
2.1
            { <, o, m, d, s }
22
23
             { >, oi, mi, di, si }
24
             { <, o, m, di, fi }
25
             \{ > , oi, mi, d, f \}
26
             { o, oi, d, di, s, si, f, fi, e }*
27
             { <, m, d, di, o, oi, f, fi, s, si, e }
28
             \{ >, mi, di, d, oi, o, fi, f, si, s, e \}
29
             \{<,>,m, mi, di, d, oi, o, fi, f, si, s, e\}*
```

Table29 is generated by our program implemented based on the following algorithms.

```
{\bf Algorithm}: Relcomp
```

Input: $rs_1 \in 29RelSet$, $rs_2 \in 29RelSet$

Output : $rs \in 29RelSet$ Preconditions : true

```
Postconditions: true 
Steps: 
 1. rs = \bigcup \forall r_1 \in rs_1, \forall r_2 \in rs_2 \quad \bullet (r_1, r_2) \in rs_1 \times rs_2

Table 13 (r_1, r_2)
```

In Function RelComp, the reasonable set computed must be the union of all possible combinations of the pair of relations obtained from the two input relation sets, name rs_1 and rs_2 . The function uses a table lookup function to obtain a set of relations.

 ${\bf Algorithm}: Compute Table 29$

Input: Table 13

```
Output: Table29
Preconditions: true
Postconditions: relation composition is closed under I
Steps:

Construct a set of 13 atomic sets from the 13 relations, assuming that this set is called I, which is an index set for table look up.
Let Table29(i, j) = Table13(i, j), i ∈ I, j∈ I
∀ Table29(i, j), i∈ I, j∈ I, do
if k = Table29(i, j) ∉ I then
I: if k = Table29(i, j)
1: if I = I ∪ Table29(i, j)
1: if I = I ∪ Table29(i, j)

3.1.2: ∀ m ∈ I, do
3.1.2.1 Table29(k, m) = Relcomp(k, m)
```

 $3.1.2.2 \ Table 29(m, k) = Relcomp(m, k)$

Table 1: The Temporal Transitive Closure Table

01 01 2 02 29 0 03 01 0 04 24 2 05 01 2 06 24 0 07 01 2 08 24 0 09 01 0	2 25 2 03 3 26 3 16 2 15 3 16 2 15 2 03 2 15	02 29 04 24 23 01 02 24	25 22 14 18 26 01 15	02 25 17 26 19 16	25 01 14 01 14 01	02 02 17 17 02 20	25 03 14 05 15	02 25 04 14	02 03 17 16	02 22 04	02 03 04	25 29 14	25 25	25 22	02	29	02	02	25	29	02	29	25	25	29 29	25 29	29 29
03 01 0 04 24 2 05 01 2 06 24 0 07 01 2 08 24 0 09 01 0	2 03 3 26 3 16 2 15 3 16 2 15 2 03 2 15	29 04 24 23 01 02 24	22 14 18 26 01 15	25 17 26 19 16	01 14 01 14 01	02 17 17 02 20	03 14 05 15	25 04 14	03 17 16	22 04	03 04	29 14	25	22											29	29	29
04 24 2 05 01 2 06 24 0 07 01 2 08 24 0 09 01 0	3 26 3 16 2 15 3 16 2 15 2 03 2 15	04 24 23 01 02 24	14 18 26 01 15	17 26 19 16	14 01 14 01	17 17 02 20	14 05 15	04 14	17 16	04	04	14			29	22	25	2.2	25	22	29	29	25	29			
05 01 2 06 24 0 07 01 2 08 24 0 09 01 0	3 16 2 15 3 16 2 15 2 03 2 15	24 23 01 02 24	18 26 01 15	26 19 16	01 14 01	17 02 20	05 15	14	16				26														
06 24 0 07 01 2 08 24 0 09 01 0	2 15 3 16 2 15 2 03 2 15	23 01 02 24	26 01 15	19 16	14 01	02 20	15			18	Λ5																
07 01 2 08 24 0 09 01 0	3 16 2 15 2 03 2 15	01 02 24	01 15	16	01	20		19	nΖ																		
08 24 0 09 01 0	2 15 2 03 2 15	02 24	15				07																		-		
09 01 0	2 03 2 15	24		02	21																						
	2 15		12																								
10 24 0																											
11 01 0																											
12 01 2																											
13 01 0																											
14 24 2																											
15 24 0																											
16 01 2																											
17 24 2 18 01 2																											
19 29 0																											
20 01 2																											
21 24 0																											
22 01 2																											
23 29 2																											
24 24 2																											
25 29 0																											
26 24 2	3 26	29	27	28	24	23	26	28	26	27	26	29	28	27	29	27	28	27	28	27	29	29	28	29	29	29	29
27 24 2																											
28 29 2																											
29 29 2																											

Algorithm ComputeTable29 adds new relation sets computed by RelcCmp to the index set I, and computes the new elements of Talble29. There are

 $C(13, 0) + C(13, 1) + C(13, 2) + ... + C(13, 13) = 2^{13}$ possible elements of I. However, from the computation of algorithm ComputeTable29, the cardinality of I results in 29. Based on this result, we argue that, for an arbitrary pair of temporal intervals, the possible relations between them must be an element of set I. Using Table29, when composing temporal relations, the set union operation is replaced by a table look up operation. Therefore, the time complexity of relation composition is reduced. The cost of memory used in Table29 is tolerable.

Definition 3.1: Given a nonempty set S = TemporalTuple, o is a binary operation on S, o is the temporal relation transitive function, the domain is $S \times S$, the codomain is $S \times S$, i.e. o: $S \times S \rightarrow S$. This mapping is sometimes called a law of composition.

Definition 3.2: To combine S and binary operation o is an algebraic system and was denoted by S, S, S.

Theorem 3.1: Let < S, o > be a temporal algebraic system, and S be a set with a law of composition, then < S, o > is closed.

Proof: Since function $o: S \times S \rightarrow S$, and S is equal to 29RelSet, the function is closed to 29RelSet.

Theorem 3.2: Let $\langle S, o \rangle$ be a temporal algebraic system, and S be a set with a law of composition, then all $a \in S$, exists $b \in S$, such that $a \circ b = b \circ a = (A, \{e\}, A)$, b is called inverse of a.

Proof: Assuming that a = (A, rs, B), where A, and B are interval names, and rs is a temporal relation set. We want to find a rs^{-1} for each rs. The following table shows the inverse relation sets rs^{-1} for each rs:

Inverse Relation Sets

rs	rs ⁻¹	rs	rs ⁻¹
1	2	18	19
3	4	20	20
5	6	21	21
7	8	22	23
9	10	24	25
11	12	26	26
13	13	27	28
14	15	29	29
16	17		

There are five relation sets which are the inverse of themselves (i.e., the one marked with a subscript "*" in the 29 relation sets). Since each relation set has its inverse, for an arbitrary a = (A, rs, B), we can always

find $b = (B, rs^{-1}, A) \in S$, such that $a \circ b = b \circ a = (A, \{e\}, A)$

Theorem 3.3: Let $\langle S, o \rangle$ be a temporal algebraic system, and S be a set with a law of composition, then $\langle S, o \rangle$ has an unique identity $(A, \{e\}, A)$. i.e. all $a \in S$, $a \circ (A, \{e\}, A) = a = (A, \{e\}, A) \circ a$.

Proof: To prove the identity of function o, we need to show that

 $\forall tt \in TemporalTuple \bullet$ $a \circ a^{-1} = (A, \{e\}, A) \land a^{-1} \circ a = (A, \{e\}, A) \land$ $a \circ (A, \{e\}, A) = a \land (A, \{e\}, A) \circ a = a$

From the table lookup of Table29, we can easily verify that $\forall rs \in 29RelSet \bullet rs \circ \{e\} = rs \land \{e\} \circ rs = rs$. It is clear that $\forall a \in TemporalTuple \bullet a \circ (A, \{e\}, A) = a \land (A, \{e\}, A) \circ a = a$. Due to Theorem 3.2, and the inverse relation sets table given above, we can look at Table29 for the composition of each pair of rs and rs^{-1} , as well as for each pair of rs^{-1} and rs.

Theorem 3.4: Let $\langle S, o \rangle$ be a temporal algebraic system, and S be a set with a law of composition, then $\langle S, o \rangle$ is associative. i.e. all $a, b, c \in S$, $(a \circ b) \circ c = a \circ (b \circ c)$.

Proof: Let L be an ordered list of relation sets obtained from I according to the order given in the 29 relation set table (i.e., L = (1, 2, 3,..., 29)). We further define L^2 to be an ordered list of elements obtained from Table29 according to the row major order. L^2 has 841 (i.e., 29^2) elements. We can easily compute a table $T_{29\times841}$ from L and L² by:

 $\forall X, Y, Z : Temporal Tuple \bullet$ $X = (A, rs_1, B) \land rs_1 = L(i) \land 1 \le i \le 29 \land Y = (B, rs_2, C) \land rs_2 = L^2(j) \land 1 \le j \le 841 \land Z = (A, rs_3, c) \land rs_3 = T(i, j) \Leftrightarrow X \circ Y = Z$

There are 24389 (i.e., 29 * 841) elements in table T. Similarly, we can compute another table from L^2 and L, named $T'_{841\times29}$. Assuming that L^3 is an ordered list of elements obtained from T according to the row major order. And L'^3 is a similar list obtained from T'. For an arbitrary i, $1 \le i \le 24389$, if L^3 (i) is the relation set of $X \circ (Y \circ Z)$, then L'^3 (i) is the relation set of $(X \circ Y) \circ Z$. We can verify that $L^3 = L'^3$. An implemented program shows the result holds.

4 Maintaining Time Constraints for Long Distances

Based on *Table29*, we propose a set of algorithms, using a directed graph, for fast temporal relation compositions. These algorithms can be used to compute the binary relation between an arbitrary pair of

intervals. User edge conflicts are eliminated and derived edges and cycles without conflict are added.

If there is a conflict cycle in the original reduced relation domain, the algorithm eliminates that conflict first by altering the user to select a reasonable relation to replace the original one. This is why the resulting graph may contain some new user edges (i.e. UE'). This conflict elimination is achieved by invoking the EliminateConflict algorithm. Suppose G is a graph of the reduce relation domain, and GV and GE are the vertex set and edge set of G, respectively. The algorithm computes derived edges based on user edges. The reason of using the user edges is that these edges contain the minimal and sufficient information of what the user wants.

```
Algorithm: ComputeRD1
Input : G = (GV, GE)
Output: K_n = (K_n V, K_n E)
Preconditions: true
Postconditions: GV = K_n V \wedge GE \setminus UE \cup UE' \subseteq K_n E
Steps:
1: G = EliminateConflicts(G)
2: K_n = G \wedge pl = 2
3: repeat until |K_nE| = |K_nV| * (|K_nV| -1)/2
    3.1: for each e = (a, b) \land e \notin K_n E \land a \in
          K_nV \wedge b \in K_nV \bullet
         there is a path of user edges from a to b, with
         path length = pl
    3.2 : suppose ((n_1, n_2), (n_2, n_3), ..., (n_{k-1}, n_k))
          is a path with a = n \land b = n \land k = pl + 1
    3.3 : set e.rs = Table 29 ((a, n_{k-1}).rs, (n_{k-1}, b).rs)
    3.4: K_nE = K_nE \cup \{e\}
    3.5: pl = pl + 1
```

The first algorithm, ComputeRD1, starts from taking each path of user edges of length 2, and computes a derived edge from that path. The insertion of edge e = (a, b) results a cycle, but no conflict. The reasonable set of edge e (i.e., e.rs) is computed from two edges, (a, n_{k-1}) and (n_{k-1}, b) , which are user edges or derived edges. Since we increase the path length, pl, of the path of user edges one by one. The derived edge (a, n_{k-1}) (or user edge, if pl = 2) must have been computed in a previous interaction. The algorithm repeats until all edges are added to the complete graph K_n , which contains n*(n-1)/2 edges.

```
Algorithm: Eliminate Conflicts

Input: G = (GV, GE)

Output: G' = (GV, GE)

Preconditions: G contains only user edges \land G' = G

Postconditions: G' = G, but the reasonable sets of edges in G' may be changed.

Steps:

1. for each P = ((n_1, n_2), (n_2, n_3), ..., (n_{k-1}, n_k)) in G' with n_1 = n_k \land k > 3
```

```
1.1 : for each i, 1 \le i \le k-2

1.1.1 : set (n_i, n_{i+2}).rs = Table29 ((n_i, n_{i+1}).rs, (n_{i+1}, n_{i+2}).rs)

1.2 : rs = Table29 ((n_k, n_{k-2}).rs, (n_{k-2}, n_{k-1}).rs)

1.3 : if (n_k, n_{k-1}).r \notin rs then

1.3.1 : ask user to choose a r' \in rs

1.3.2 : set (n_k, n_{k-1}).r = r'
```

Considering the five user edges, the algorithm computes derived edges until the last edge is added to K_n : User edges:

```
(A, B) = { < } = [1]
(B, C) = { m } = [7]
(C, D) = { d } = [3]
(C, E) = { s } = [9]
(F, D) = { < } = [1]
```

Derivation based on user edges:

```
1. Path Length = 2
   (A, C) = (A, B) \circ (B, C) = [1] \circ [7] = [1] = \{ < \}
   (B, D) = (B, C) \circ (C, D) = [7] \circ [3] = [16] = \{0, d, s\}
   (C, F) = (C, D) \circ (D, F) = [3] \circ [1]^{-1} = [3] \circ [2] = \{ > \}
   (D, E) = (D, C) \circ (C, E) = [4] \circ [9] = [14] = \{0, di, fi\}
   (B, E) = (B, C) \circ (C, E) = [7] \circ [9] = [7] = \{ m \}
2. Path Length = 3
   (A, E) = (A, B) \circ (B, C) \circ (C, E) = (A, C) \circ (C, E)
        = [1] \circ [9] = [1] = \{ < \}
   (A, D) = (A, B) \circ (B, C) \circ (C, D) = (A, C) \circ (C, D)
             = [1] \circ [3] = [22] = \{ <, o, m, d, s \}
   (B, F) = (B, C) \circ (C, D) \circ (D, F) = (B, D) \circ (D, F)
         = [16] o [1]^{-1} = [23] = { >, oi, mi, di, si }
   (E, F) = (E, C) \circ (C, D) \circ (D, F) = (E, D) \circ (D, F)
         =[14]^{-1} \circ [2] = [15] \circ [2] = [2] = \{ > \}
3. Path Length = 4
   (A, F) = (A, B) \circ (B, C) \circ (C, D) \circ (D, F)
         = ((A, B) \circ (B, C)) \circ ((C, D) \circ (D, F))
         = (A, C) \circ (C, F) = [1] \circ [2] = [29]
         = \{ <, >, d, di, o, oi, m, mi, f, fi, s, si, e \}
```

5 Extending algorithms to N-Dimensional Spatial Relations

Let rs denote a set of 1-D temporal interval relations (i.e., $rs \in 29Relset$). The relation composition table discussed in [1] can be refined (e.q., make each relation as an atomic set of that relation) to a function maps from the Cartesian product of two rs to a rs. Assuming that f^l is the mapping function interpreting Allen's table, we can compute f^2 , the relation composition function of 2-D objects, and f^3 , the one for 3-D objects, from f^l . There are 13 relations for 1-D objects. A conjunction of two 1-D relations, which denotes a 2-D relation, has 13^2 variations. Similarly, there are 13^3 3-D relations.

```
f^{l} = 29RelSet \times 29RelSet \rightarrow 29RelSet

f^{2} = 29RelSet \times 29RelSet \times 29RelSet \times 29RelSet \rightarrow

29RelSet \times 29RelSet

f^{3} = 29RelSet \times 29
```

```
29RelSet × 29RelSet → 29RelSet × 29RelSet ×
           29RelSet
     where 29RelSet \times 29RelSet \in \{ \{ \le \} \times \{ \le \}, \{ \le \} \}
     \{>\}, \dots, \{=\} \times \{=\} \}
     29RelSet \times 29RelSet \times 29RelSet \in \{ \{<\} \times \{<\} \},
     \{<\}\times\{<\}\times\{>\},..., \{=\}\times\{=\}\times\{=\}\}
    Functions f^2 and f^3 are computed according to the
following formulas:
    \forall i_1 \times j_1, i_2 \times j_2 \in P (29RelSet \times 29RelSet)
           f^{2}(i_{1} \times j_{1}, i_{2} \times j_{2}) = \prod f^{1}(i_{1}, i_{2}) \times f^{1}(j_{1}, j_{2})
    \forall i_1 \times j_1 \times k_1, i_2 \times j_2 \times k_2 \in P(29RelSet \times 29RelSet)
           \times 29RelSet)
    f^{3}(i_{1} \times j_{1} \times k_{1}, i_{2} \times j_{2} \times k_{2}) = \prod f^{I}(i_{1}, i_{2}) \times f^{I}(j_{1}, j_{2})
           \times f^{I}(k_{I}, k_{2})
   where \prod A \times B = \{ a \times b \mid \forall a \in A, b \in B \}
   \prod A \times B \times C = \{ a \times b \times c \mid \forall a \in A, b \in B, c \in C \}
```

The functions are implemented as table mappings. Table generated by the above formula are stored in memory to reduce run-time computation load.

6 The Applications

Spatio-Temporal relations can be used in many multimedia related applications. Using our proposed algorithms in this paper, inference rules can be generalized to generate a better presentation.

As long as the spatial relations of objects are decided, the algorithm can compute the location of each presentation resource in a window. However, parameters to spatial relations need to be added to precisely specify relative screen coordinates.

Moreover, Spatio-Temporal relations can be used to compose objects in multimedia documents [8]. Our mechanism can be incorporated with an objectoriented mechanism [7] for the construction of reusable multimedia documents. Multimedia resources are not used along usually. Instead, they have some degree of associations. For instance, a motion picture resource is synchronized with a MIDI song as its background music. Similarly, multimedia resources, when they are presented, may have some default relative positions. Spatial relations can be used in this case. In an appli-cation, the presentation system can use our algorithms to compute the schedule of a presentation. Also, in natural language processing, temporal intervals are used to model the timing of events. Our algorithms thus can be used in constructing semantics of sentences. We believe that, spatiotemporal relations can be used in many related applications for maintaining time constraints.

7 Conclusions

The main contributions of this paper is in building the algebra system of spatio-temporal interval relations and the set of enhanced mechanism for spatiotemporal relation composition. These algo-rithms deal with an arbitrary number of objects in an arbitrary ndimensional space. We propose many properties of temporal interval relations and prove the correctness of these properties. We also argue that, many interesting researches in multimedia applications can benefit from using these spatio-temporal relations and our algorithms.

The algorithm proposed in this paper can be used in other computer applications. We hope that, with our analysis and algorithms, the knowledge underlying temporal interval relations can be used in many computer applications, especially in distributed multimedia computing and networking.

References

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