

Maintaining Knowledge about Temporal Intervals

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Temporal Knowledge

- Needed to temporal reasoning problems
- Outdated data can be important:
 - “Which employees worked for us last year?”
- We need to represent temporal information

Examples from AI

- Planning
 - robot's actions on the world must be taken into account to ensure that a plan will be effective
- Natural language processing
 - Temporal information in sentences are extracted from sentences to be able to answer queries

Temporal Intervals

- Temporal interval is a primitive notion
 - Relationships between temporal intervals
 - Based on hierarchy
 - Constraint propagation techniques
 - Temporal knowledge is relative
 - Cannot be described by a date

Contents

- Interval-based systems vs. point-based systems
- Interval-based deduction technique based on constraint propagation

Temporal representation characteristics

- Should allow imprecision
 - A before B
- Uncertainty of information
 - we don't know the exact relationship between A and B but some constraints on how they could be related are known
- Grain of reasoning
 - Days, years... nanoseconds!
- Persistence
 - Default reasoning: "I parked my car this morning, it should be there now"

Temporal systems approaches

- State space approaches
- Date line systems
- Before/after chaining
- Formal models

State space approaches

- Only for simple problem solving tasks
- A state is a description of the world
 - Actions are functions mapping between states
- Previous states:
 - Can be retained (better representation)
 - Can be deleted: present state only (efficiency)
- Do provide a notion of persistence

Database systems

- Each fact is indexed by a date
 - Date: representation of a time such that the temporal ordering between 2 dates can be computed by simple operations
 - Integers, calendar dates
- Unfortunately we cannot assign a precise date for every event
 - Fuzzy extensions
- A and B not together
 - Time intervals make the harder

Before/After chaining

- Capture relative temporal information quite directly
- Computational problems:
 - When the amount of temporal information grows
 - Search problems (long chains)
 - Space problems (if all possible relationships are precomputed)
- Allan's work is based on B/A chaining

Formal models

- From philosophy
- From AI
 - Situation calculus
 - McDermott work
- All of them are point-based theories
 - Time intervals are created from time points

Time points vs. time intervals: sentences

Precise point in time at which the finding event occurred

- We found the letter at twelve noon

■ Time interval

- We found the letter yesterday

■ Temporal relation between time intervals

- We found the letter while John was away
- We found the letter after we made the decision

Time points vs. time intervals: decomposition

- “finding the letter”
 - “looking at spot X where the letter was”
 - “realizing that it was the letter you were looking at”
- And then decompose “realizing...”
- The times we consider are times of events
 - We can decompose events
 - Time point seems to be not useful
 - Time points can be useful if they are “very small intervals”

Time points vs. time intervals: endpoints

- A situation where a light is turned on
- Two interval times: light off, light on
 - (off) (on)
 - [off] [on]
- An “artificial” solution
 - [off) [on)
- Time based on points does not correspond to our intuitive notion of time
 - intervals as primitive!

Time points vs. time intervals: intervals as primitive

- Definition of intervals using endpoints
 - An interval t is an ordered pair of points with the first point less than the second
 - $(t-, t+)$
 - $t- < t+$

Interval Relations defined by endpoints

<i>Interval relation</i>	<i>Equivalent Relations on Endpoints</i>
$t < s$	$t+ < s-$
$t = s$	$(t- = s-) \ \& \ (t+ = s+)$
$t \text{ overlaps } s$	$(t- < s-) \ \& \ (t+ > s-) \ \& \ (t+ < s+)$
$t \text{ meets } s$	$t+ = s-$
$t \text{ during } s$	$((t- > s-) \ \& \ (t+ = < s+)) \text{ or } ((t- >= s-) \ \& \ (t+ < s+))$

13 relations

Relation	Symbol	Sym inverse	Pict Example
X before Y	<	>	XXX YYY
X equal Y	=	=	XXX YYY
X meets Y	m	mi	XXXYYY
X overlaps Y	o	oi	XXX YYY
X during Y	d	di	XXX YYYYYYY
X starts Y	s	si	XXX YYYYYYY
X finishes Y	f	fi	XXX YYYYYYY

Network representation

- Nodes represent individual intervals
- Arcs are labeled to indicate the possible relationship between two intervals (their nodes)
- Uncertainty → all possible cases are entered on the arc
- Note: 13 possibilities are mutually exclusive
 - No ambiguity in this notation

Representing knowledge of temporal relations in a Network

- i during j $N_i - (d) \rightarrow N_j$
- i during j or
 i before j or
 j during i $N_i - (< d d_i) \rightarrow N_j$
- $(i < j)$ or $(i > j)$ or
 i meets j or
 j meets i $N_i - (< > m m_i) \rightarrow N_j$

Basic algorithm

Given $R1, R2$ arc labels (two sets of relations)

Constraints($R1, R2$)

$C \leftarrow \varepsilon$

for each $r1$ in $R1$

for each $r2$ in $R2$

$C \leftarrow C \text{ union } T(r1, r2);$

return C ;

Basic algorithm (2)

To Add $R(i,j)$

Add $\langle i,j \rangle$ to queue ToDo

While ToDo is not empty do

 Get next $\langle i,j \rangle$ from ToDo

$N(i,j) \leftarrow R(i,j)$

For each node k **such that** Comparable(k,j) do

$R(k,j) \leftarrow N(k,j) \cap \text{Constraints}(N(k,i), R(i,j))$

 If $R(k,i)$ contained in $N(k,i)$

 then add $\langle k,i \rangle$ to ToDo

For each node k **such that** Comparable(i,k) do

$R(i,k) \leftarrow N(i,k) \cap \text{Constraints}(N(i,k), R(i,k))$

 If $R(i,k)$ contained in $N(k,i)$

 then add $\langle i,k \rangle$ to ToDo

end while

An example

- “John was not in the room when I touched the switch to turn on the light”
 - S is the time of touching the switch
 - L is the time the light was on
 - R is the time that John was in the room
- The network:
 - $R \leftarrow (< m \text{ mi } >) - S - (o \text{ m}) \rightarrow L$

An example (2)

- When we insert the second fact ($R \leftarrow S$) we compute:

$$T(o_i, <) = (< o m di fi)$$

$$T(o_i, m)$$

...

$$T(m_i, >) = (>)$$

$$R \leftarrow (< > o oi m di s si fi =) - L$$

An example (3)

- And then: “But John was in the room later while the light went out”

L overlaps, starts or during R

$L - (o \ s \ d) \rightarrow R$

An example (4)

- We eliminate the impossible relationship d (by intersection)
- We propagate the effects of the constraints through the network
 - A new constraint between S and R can be calculated using the path:
$$S - (om) \rightarrow L - (os) \rightarrow R$$

We find $S - (< o m) \rightarrow R$ and intersecting $S - (< m) \rightarrow R$

An example (5)

- Finally we have:

$$R \leftarrow (< m) - S - (o m) \rightarrow L$$

$$R \leftarrow \text{——} (o s) \text{——} L$$

$$R \leftarrow (< m) - S$$

(John entered the room either after I touched the switch or at the same time that I finished touching the switch)

Conclusion

- Good balance between expressive power and efficiency of its deductive engine
- High efficiency in time
- High space requirement solved by some extensions to the basic algorithm
 - Reference Intervals (Hierarchy among time intervals)
 - Cluster of intervals for which the temporal constraints between each pair of intervals is fully computed