Solving Minimal Constraint Networks in Qualitative Spatial and Temporal Reasoning

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Minimal networks

Complete networks in which *each* tuple in any constraint can be extended to a solution

$$x \{(1,2), (\underline{1,3})\} y, y \{(2,1), (\underline{3,2})\} z, x \{(1,1)\} z$$

Solution:
$$x = 1, y = 2, z = 1$$

Not minimal: x = 1, y = 3 can not be extended

- A minimal network is satisfiable
- Computing a solution of a minimal network is NPhard [Gottlob CP11]

– If the above problem is in P, then SAT is also in P

• How is the case in QSTR?

Outline

- Qualitative spatial and temporal reasoning
 - Qualitative calculi
 - Interval Algebra, Cardinal Relation Algebra, RCC-5/8
 - CSP in QSTR
- Main result:

Computing a solution of minimal networks in above qualitative calculi is NP-hard.

• Proof sketch

What's QSTR ?

- Qualitative Spatial and Temporal Reasoning : <u>Represent</u> and <u>reason with</u> spatial (or temporal) knowledge in a qualitative manner
 - Québec City is to the northeast of Montreal
 - Montreal is to the northeast of Toronto

- Vs. quantitative approach
 - High level
 - Closer to human cognition



Qualitative calculus

- Language in QSTR
- Dealing with a certain aspect (e.g., topology, direction, size) of space or time
- Characterized by its universe and basic relations

Universe: U

Set of spatial (or temporal) entities

Basic relations:

$$B = \{b_1, b_2, \dots, b_n\}$$

A partition of U^k (in this work k=2)

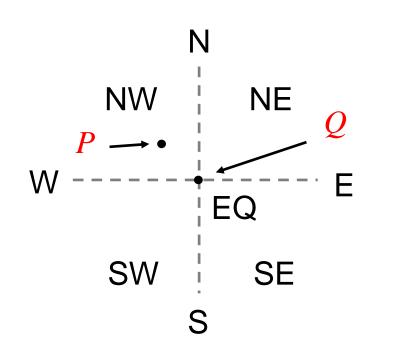
Point Algebra U: Real numbers B: {<, =, >}

Cardinal Relation Algebra

- Universe : Real plane
- Basic relations:

NW, N, NE, W, EQ, E, SW, S, SE

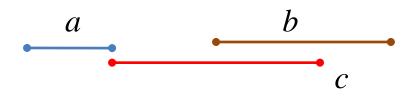
P NW Q



Interval Algebra

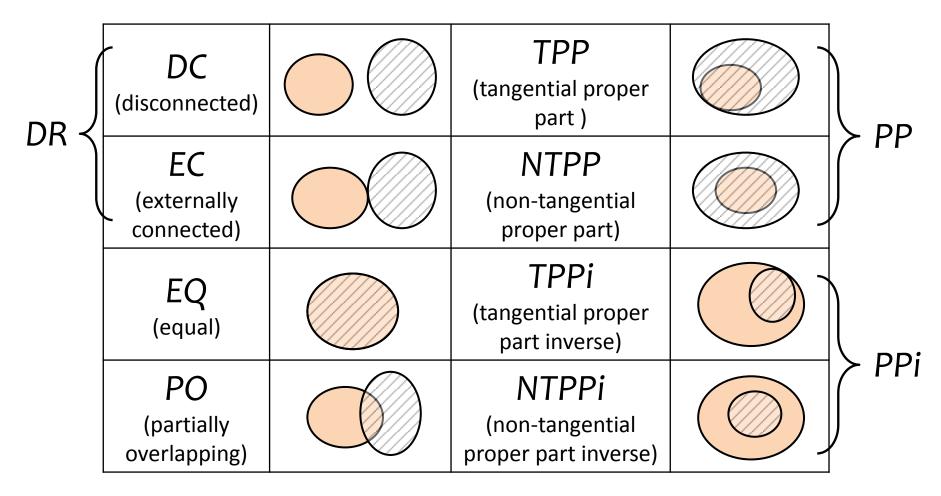
- Universe : all closed intervals $[x^-, x^+]$
- 13 basic relations

Relation	Symbol	Converse	Meaning	
before	р	pi	$x^- < x^+ < y^- < y^+$	• • •
meets	m	mi	$x^- < x^+ = y^- < y^+$	• • •
overlaps	0	oi	$x^- < y^- < x^+ < y^+$	• • •
starts	S	si	$x^- = y^- < x^+ < y^+$	•
during	d	di	$y^- < x^- < x^+ < y^+$	• • ••••
finishes	f	fi	$y^- < x^- < x^+ = y^+$	• • ••••••••••••••••••••••••••••••••••
equals	eq	eq	$x^- = y^- < x^+ = y^+$	⊨



RCC-8 and RCC-5

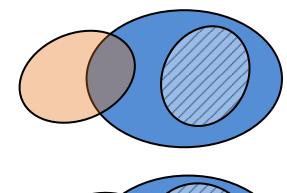
Universe: *plane regions*



CSP in QSTR

- Qualitative calculus: a constraint language
 - The domain of each variable is the universe
 - The relation in each constraint is a set of basic relations
- Examples:
 - $-x \{DC, EC\} y, y NTPP z, x PO z$
 - Satisfiable, minimal
 - $v_1 < v_2, v_2 \{<,=\} v_3, v_1 \{<,>\} v_3$
 - Satisfiable, not minimal
 - $v_1 > v_3$ can not be satisfied

Complexity of solving minimal networks (in different qualitative calculi) in QSTR?



Minimal networks in QSTR

- For CRA, IA, RCC-5/8, the answer is NP-hard
- Though Gottlob's approach for proving the NP-hardness is followed, our result is not implied directly:

Domains of variables

free VS. assumed (infinite)

Constraints

free VS. restricted (basic relations as tuples)

Proof technique (Gottlob 11)

Solving a minimal network is NP-hard:

If there is a polynomial algorithm \mathcal{A} that computes a solution of a minimal network, then some NP-hard problem \mathcal{P} can be solved in polynomial time by an algorithm based on \mathcal{A} ...

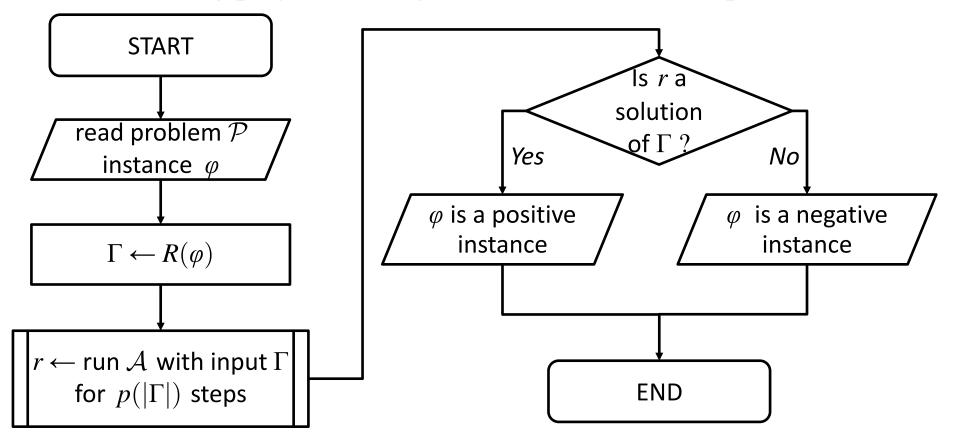
Provided that there exists a polynomial mapping R from P to CSP, such that for instance φ in problem P,

(*) φ is positive iff $R(\varphi)$ is a minimal network; and (**) φ is negative iff $R(\varphi)$ is an unsatisfiable network.

R is a reduction, therefore, deciding the minimality of CSP is NP-hard.

Proof technique (Gottlob 11)

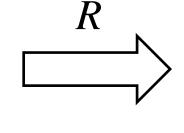
(1) *A* computes a solution of a minimal network in *p*(.) time
(2) φ is positive (negative) iff *R*(φ) is minimal (unsatisfiable)
The following polynomial algorithm solves NP-hard problem *P*



To prove the NP-hardness...

- Find an NP-hard problem \mathcal{P} , and
- A reduction R from \mathcal{P} to the target CSP, s.t.
 - Positive instances mapped to minimal networks
 - Negative instances mapped to unsatisfiable networks.

Symmetry of instances of \mathcal{P}



Minimality of target CSP networks

Symmetric SAT

A SAT instance φ is symmetric if either φ is unsatisfiable, or for any an assignment π , π satisfies φ implies that assignment $\overline{\pi}$ is also satisfying, where π assigns each propositional variable p the opposite truth value to $\overline{\pi}(p)$.

Lemma

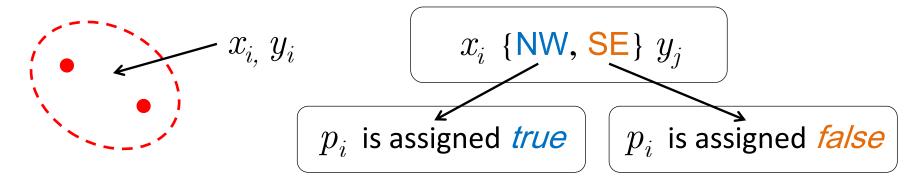
A SAT instance φ can be transformed in polynomial time into a symmetric SAT instance φ^* , preserving satisfiability.

 $\varphi = (p_1 \lor \neg p_2) \land (p_2 \lor p_3)$ $\varphi^* = (p_1 \lor \neg p_2 \lor q) \land (p_2 \lor p_3 \lor q) \land (\neg p_1 \lor p_2 \lor \neg q) \land (\neg p_2 \lor \neg p_3 \lor \neg q)$

Symmetric SAT is NP-hard

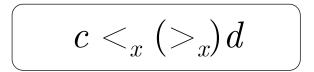
Proof sketch, Cardinal Relation Algebra

Propositional variable $p_i \implies$ spatial variables x_i and y_i



Literal $l \implies$ spatial variables c and d

l is assigned *true* (*false*) : c is to the left (right) of d



Proof sketch, Cardinal Relation Algebra

- For clause c contains literals l₁, ..., l_t
 We have spatial variables c₁, d₁, ..., c_t, d_t
- We impose constraints such that c_{k+1} , d_k are on the same vertical line (c_{t+1} considered as c_1) $c_{k+1} =_r d_k$
- All literals are assigned false :

$$c_1 >_x d_1 =_x c_2 >_x d_2 =_x \dots =_x c_t >_x d_t =_x c_1$$

- Symmetry of the SAT instances also forbids the case that all literals are assigned true.
- The constructed CRA network is minimal if SAT instance is satisfiable.

Interval Algebra

- An interval [x, y] corresponds to a point (x, y)
- Translate previous reduction

NW	Ν	NE	W	EQ	Ε	SW	S	SE
di	si	oi	fi	eq	f	0	S	d

RCC-5/8

k-supersymmetric **SAT** [Gottlob 11]:

A SAT instance φ is *k*-supersymmetric if either φ is unsatisfiable, or arbitrary partial truth value assignment over *k* variables can be extended to a satisfying assignment of φ .

Lemma

A SAT instance φ can be transformed in polynomial time into a symmetric and *k*-supersymmetric SAT instance φ^* , preserving satisfiability.

A more delicate reduction is needed... refer to the paper for details.

Conclusion

- Solving a minimal network in qualitative calculi IA, CRA, RCC-5/8 is NP-hard.
- Bi-product: deciding minimality in these qualitative calculi is NP-hard.

- Thank you for your attention !
- Questions?