

**Class Discussion on Advanced Consistency Methods (Sections 8.1-8.3)**  
**by Yaling Zheng (March 24-April 2)**  
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**1 Overview**

Until this class we have seen variable-based consistency algorithms. In this chapter, Yaling discussed relation-based consistency enforcing algorithms.

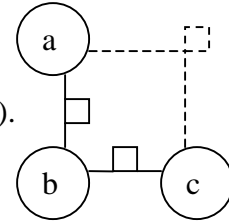
**8.1 Relational consistency**

**Relational-arc-consistency (RAC):** given a network  $R$  (CSP) and a relation  $R_S$  over a set  $S$ ,  $R_S$  is said to be RAC relative to a variable  $x \in S$  iff any instantiation of variables in  $S - \{x\}$  could be extended to a value in  $D_x$  that satisfies  $R_S$ :

$$\rho(S - \{x\}) \subseteq \pi_{S - \{x\}}(R_S \bowtie D_x)$$

A relation  $R_S$  is RAC if it is RAC relative to every variable in  $S$ .  
 Finally, a CSP is RAC iff every constraint in CSP is RAC.

Example:  $\{a, b, c\}$  are variables,  $R_{ab}$  and  $R_{bc}$  are relations (constraints).  
 And newly formed constraint (shaded rectangle with shaded edges)  
 $R_{ac}$  is RAC relative to 'b'. Where, 'b' is a common variable.



**Relational-path-consistency (RPC):** given a network  $R$  (CSP) and relations  $R_S$  and  $R_D$ ,  $R_S$  and  $R_D$  are said to be RPC relative to a variable  $x \in S \cap T$  iff any instantiation of variables in

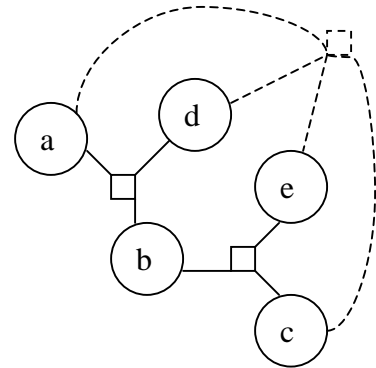
$A = S \cup T - \{x\}$  could be extended to a value in  $D_x$  that satisfies both  $R_S$  and  $R_D$ :

$$\rho(A) \subseteq \pi_A(R_S \bowtie R_T \bowtie D_x)$$

A pair of relations  $R_S$  and  $R_D$  is RPC iff it is RPC relative to every variable in  $S \cap T$ .

Finally, a CSP is RPC iff every pair of its relations is RPC.

Example:  $\{a, b, c, d, e\}$  are variables,  $R_{abd}$  and  $R_{bce}$  are relations (constraints). And newly formed constraint  $R_{adce}$  is RPC relative to 'b', where 'b' is a common variable.



From class discussion:

- If a graph is complete or triangulated RPC will guarantee global consistency.
- We are inferring a new relation between variables that are not related.
- In binary-CSP we use relations to update domains, while in relational-CSP we use a domain of one variable to update domain of other variables.

**Relational-m-consistency (R-m-C):** given a network  $\mathbf{R}$  (CSP) and relations  $R_{S_1}, \dots, R_{S_m}, R_{S_1}, \dots, R_{S_m}$  are said to be R-m-C relative to a variable  $x \in \prod_{i=1}^m S_i$  iff any instantiation of variables in  $A = \prod_{i=1}^m S_i - \{x\}$  has an extension to a value in  $D_x$  that satisfies all  $R_{S_1}, \dots, R_{S_m}$  constraints:

$$\rho(A) \subseteq \pi_A(\langle\langle \prod_{i=1}^m R_{S_i} \rangle\rangle D_x)$$

Set of relations  $\{R_{S_1}, \dots, R_{S_m}\}$  is R-m-C iff it is R-m-C relative to every variable in  $\prod_{i=1}^m S_i$ .

Finally, a CSP is R-m-C iff every set of  $m$  relations is R-m-C, a CSP is strong R-m-C if it is R-i-C for every  $i \leq m$ .

### 8.1.1 Space-bound vs time-bounds

**Relational (i, m)-consistency (R(i,m)-C):** a set of relations  $\{R_{S_1}, \dots, R_{S_m}\}$  is R(i,m)-C iff for every subset (A) of size  $i$   $A \subseteq \prod_{j=1}^m S_j$ , any consistent assignment to A could be extended to  $\prod_{j=1}^m S_j - A$  that satisfies all relation in the set  $\{R_{S_1}, \dots, R_{S_m}\}$ . Similarly the CSP would be R(i,m)-C iff every set of size  $m$  relations is R(i,m)-C. In addition, a CSP is strong R(i,m)-C iff it is R(j,m)-C for every  $j \leq i$ .

From class discussion:

- Inverse arc consistency and arc consistency are identical (relational (1,1)-consistency and AC)
- Neighborhood inverse consistency was discussed (for those who are interested briefly[4], advanced paper [5])

### 8.2 Directional-relational-consistency

**Directional relational consistency (DRC):** given a network  $\mathbf{R}$  (CSP) and ordering  $d=(x_1, \dots, x_n)$ ,  $\mathbf{R}$  is m-directionally relationally consistent iff

- for every subset of constraints  $\{R_{S_1}, \dots, R_{S_m}\}$  where the latest variable in any of the constraints' scopes is  $x_l$
- for every A,  $A \subseteq \{x_1, \dots, x_{l-1}\}$ , every consistent assignment to A can be extended to  $x_l$
- simultaneously satisfies all relevant constraints in  $\{R_{S_1}, \dots, R_{S_m}\}$ .

**Complexity of DRC:**

- time complexity of  $DRC_m$  along ordering  $d$  is  $O\left(nm \cdot (2^m k^2)^{w^*(d)+1}\right)$ ,  $w^*(d)$  is width of a directional graph,  $n$  is number of variables, and  $k$  bounds the domain size
- time complexity of  $DRC_2$  is  $O\left(n(4k^2)^{w^*(d)+1}\right)$
- complexity of  $DRC_2$  is  $O\left(n \cdot e^2 \cdot t^2\right)$  where  $e$  is the number of input relations and  $t$  bounds the number of tuples in each relation

### 8.3 Domain and constraint tightness

#### *Domain tightness:*

- A strong relational 2-consistent network over bi-valued domains is globally consistent
- A strong relational  $k$ -consistent constraint network  $R=(X,D,C)$  over domains with at most  $k$  values is globally consistent

#### *Constraint tightness:*

- A constraint relation  $R_S$  of arity  $r$  is called  $m$ -tight if, for any variable  $x_i \in S$  and any instantiation  $\bar{a}$  of the remaining  $r-1$  variables in  $S-\{x_i\}$ , either there are at most  $m$  extensions of  $\bar{a}$  to  $x_i$  that satisfy  $R_S$ , or there are exactly  $|D_S|$  such extensions
- A strongly relationally  $(m+1)$ -consistent constraint network with  $m$ -tightness is globally consistent

### 8.4 Inference for Boolean theories

Resolution operation over two clauses  $(\alpha \vee Q)$  and  $(\alpha \vee \neg Q)$  returns

$$(\alpha \vee \beta) = EC_Q(\text{models}(\alpha \vee Q), \text{models}(\beta \vee \neg Q)),$$

where  $Q'$  is the union of scopes of both clauses excluding  $Q$  (there are typos in the slide #44, AND operators should be changed to OR in the second clause, and it is “and” not OR between two clauses).

The *resolution operation* could be generalized extending literals  $\alpha$  and  $\beta$  to multiple *conjunctions* of literals as long as both clauses contain a common literal (should be negated in either clause).

Relevance to the discussion: “resolution operation” could modify  $RC_2$  to solve *CNF* theories. Extended 2-composition from  $RC_2$  is replaced with resolution, and applied until no new resolvents generated.

Similarly “resolution operation” could be used to modify  $DRC_2$  into Directional-resolution ( $DR$ ). In the  $DR$  algorithm extended 2-composition replaced with resolution and instantiation is replaced with unit resolution.

Input: a Boolean network  $\phi$  in CNF form and an ordering  $d$ .

Output: A decision whether  $\phi$  is satisfiable or not.

#### **Reference:**

- [1] Yaling Zheng, *Lecture Slides*
- [2] Rina Dechter, *Constraint Processing*
- [3] Class, *Class discussions*
- [4] <http://ic.arc.nasa.gov/ic/projects/mba/cs329a/handouts/handout-8.pdf>
- [5] <ftp://ftp.cs.unh.edu/pub/csp/Papers/aaai96-nic-cde-ecf.ps.gz>