

# CompSci 275, CONSTRAINT Networks

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Consistency algorithms, part a  
Chapter 3

# Outline

- Arc-consistency algorithms
- Path-consistency and i-consistency
- Arc-consistency, Generalized arc-consistency, relation arc-consistency
- Global and bound consistency
- Distributed (generalized) arc-consistency
- Consistency operators: join, resolution, Gaussian elimination

# Consistency methods

- Approximation of inference:
  - Arc, path and i-consistency
- Methods that transform the original network into tighter and tighter representations

## Inference Algorithms will help search

**Definition 3.1.1 (partial solution)** *Given a constraint network  $\mathcal{R}$ , we say that an assignment of values to a subset of the variables  $S = \{x_1, \dots, x_j\}$  given by  $\bar{a} = (\langle x_1, a_1 \rangle, \langle x_2, a_2 \rangle, \dots, \langle x_j, a_j \rangle)$  is consistent relative to  $\mathcal{R}$  iff it satisfies every constraint  $R_{S_i}$  such that  $S_i \subseteq S$ . The assignment  $\bar{a}$  is also called a partial solution of  $\mathcal{R}$ . The set of all partial solutions of a subset of variables  $S$  is denoted by  $\rho_S$  or  $\rho(S)$ .*

# Arc-consistency

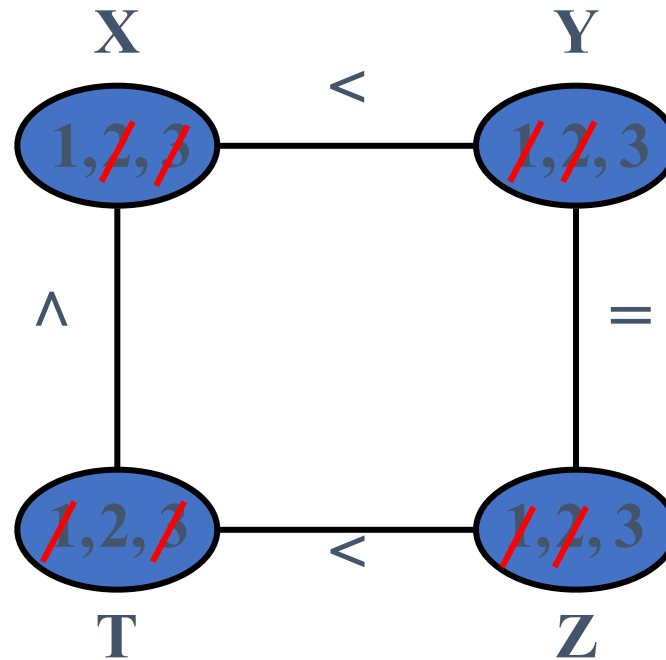
$$1 \leq X, Y, Z, T \leq 3$$

$$X < Y$$

$$Y = Z$$

$$T < Z$$

$$X \leq T$$



# Arc-consistency

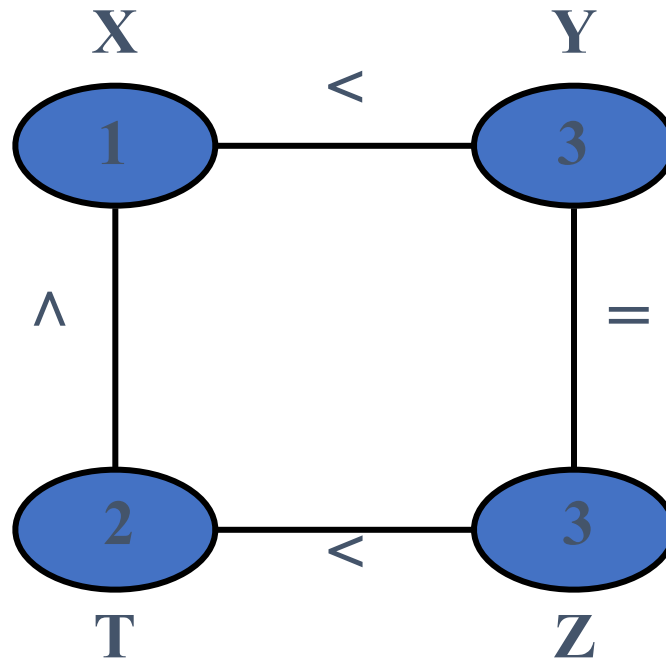
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# Arc-consistency

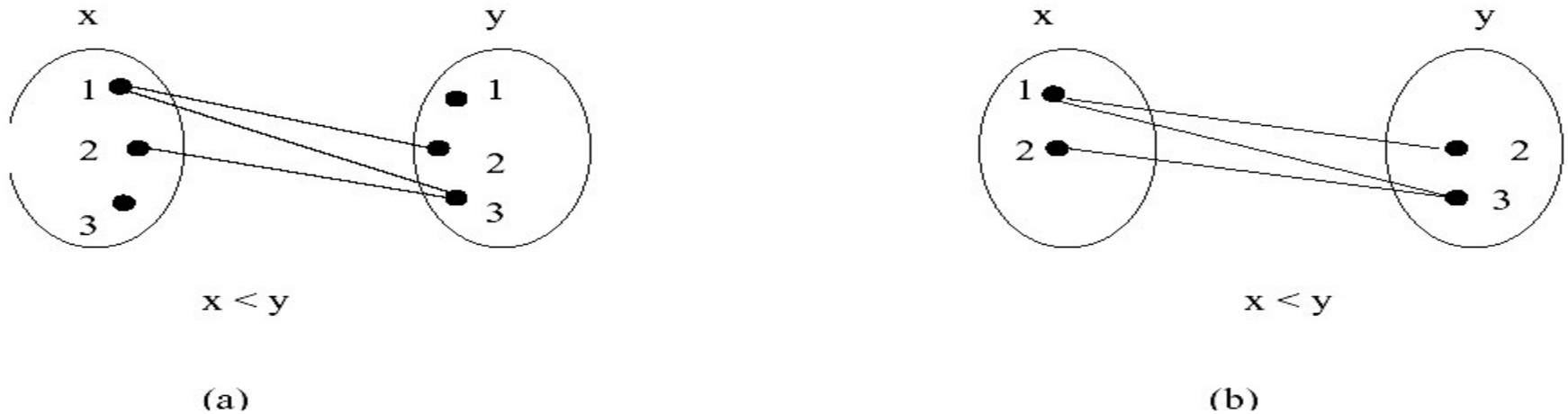


Figure 3.1: A matching diagram describing the arc-consistency of two variables  $x$  and  $y$ . In (a) the variables are not arc-consistent. In (b) the domains have been reduced, and the variables are now arc-consistent.

**Definition 3.2.2 (arc-consistency)** Given a constraint network  $\mathcal{R} = (\mathcal{X}, \mathcal{D}, \mathcal{C})$ , with  $R_{ij} \in \mathcal{C}$ , a variable  $x_i$  is arc-consistent relative to  $x_j$  if and only if for every value  $a_i \in D_i$  there exists a value  $a_j \in D_j$  such that  $(a_i, a_j) \in R_{ij}$ . The subnetwork (alternatively, the arc) defined by  $\{x_i, x_j\}$  is arc-consistent if and only if  $x_i$  is arc-consistent relative to  $x_j$  and  $x_j$  is arc-consistent relative to  $x_i$ . A network of constraints is called arc-consistent iff all of its arcs (e.g., subnetworks of size 2) are arc-consistent.

# Revise for arc-consistency

REVISE( $(x_i), x_j$ )

**input:** a subnetwork defined by two variables  $X = \{x_i, x_j\}$ , a distinguished variable  $x_i$ ,  
domains:  $D_i$  and  $D_j$ , and constraint  $R_{ij}$

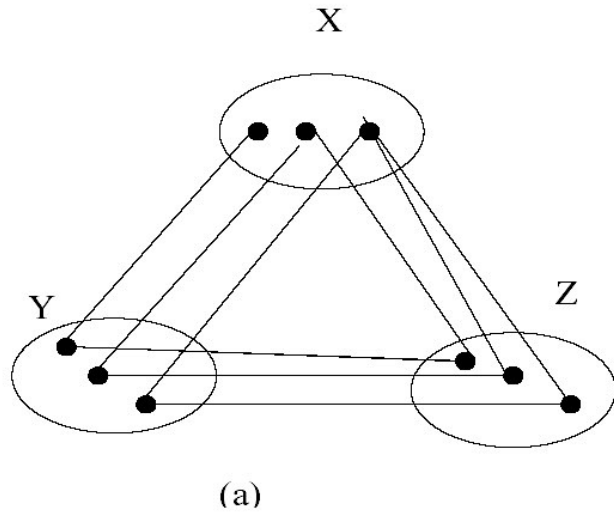
**output:**  $D_i$ , such that,  $x_i$  arc-consistent relative to  $x_j$

1. **for** each  $a_i \in D_i$
2.     **if** there is no  $a_j \in D_j$  such that  $(a_i, a_j) \in R_{ij}$
3.         **then** delete  $a_i$  from  $D_i$
4.     **endif**
5. **endfor**

Figure 3.2: The Revise procedure

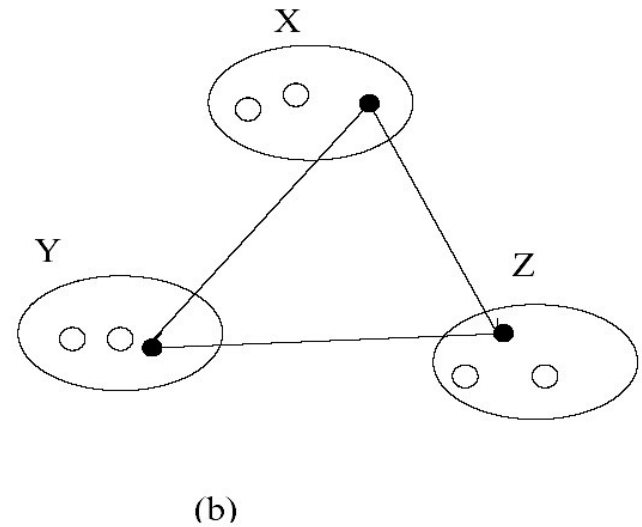
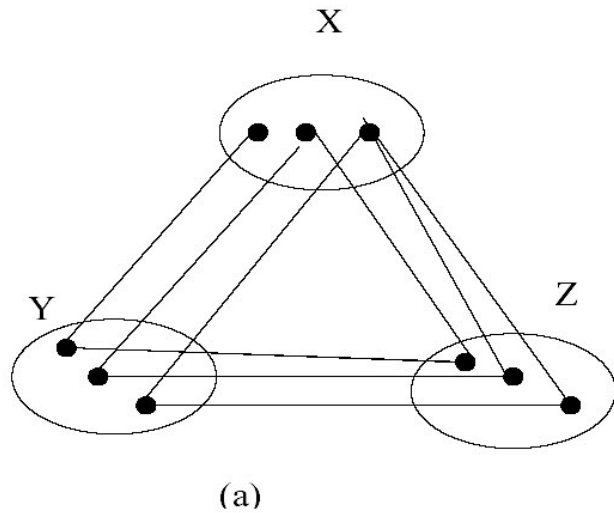


A matching diagram describing a network of constraints that is not arc-consistent (b) An arc-consistent equivalent network.





A matching diagram describing a network of constraints that is not arc-consistent (b) An arc-consistent equivalent network.



# AC-1

AC-1( $\mathcal{R}$ )

**input:** a network of constraints  $\mathcal{R} = (X, D, C)$

**output:**  $\mathcal{R}'$  which is the loosest arc-consistent network equivalent to  $\mathcal{R}$

1. **repeat**
2.     **for** every pair  $\{x_i, x_j\}$  that participates in a constraint
3.         Revise( $(x_i), x_j$ ) (or  $D_i \leftarrow D_i \cap \pi_i(R_{ij} \bowtie D_j)$ )
4.         Revise( $(x_j), x_i$ ) (or  $D_j \leftarrow D_j \cap \pi_j(R_{ij} \bowtie D_i)$ )
5.     **endfor**
6. **until** no domain is changed

Figure 3.4: Arc-consistency-1 (AC-1)

- Complexity (Mackworth and Freuder, 1986):
- $e$  = number of arcs,  $n$  variables,  $k$  values
- $(ek^2, \text{ each loop, } nk \text{ number of loops})$ , best-case =  $ek$ ,
- Arc-consistency is:

# AC-3

AC-3( $\mathcal{R}$ )

**input:** a network of constraints  $\mathcal{R} = (X, D, C)$

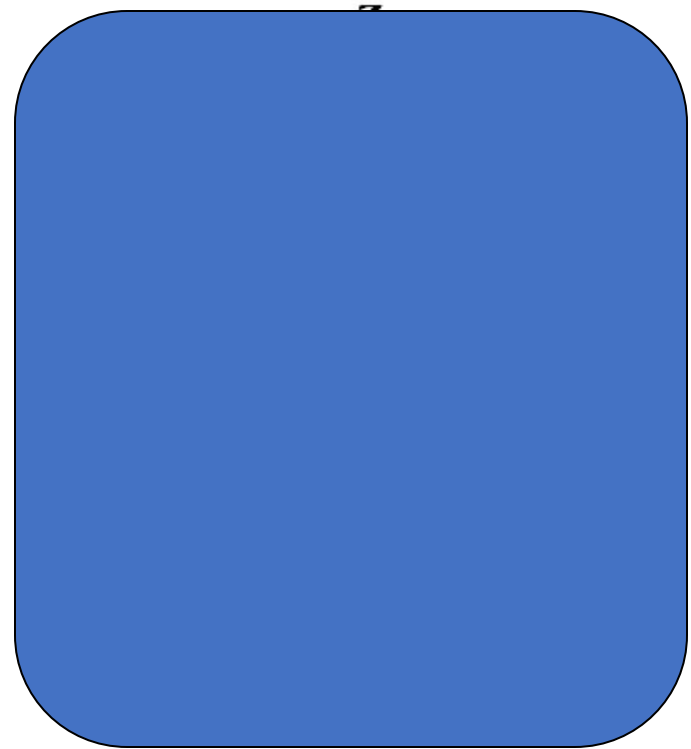
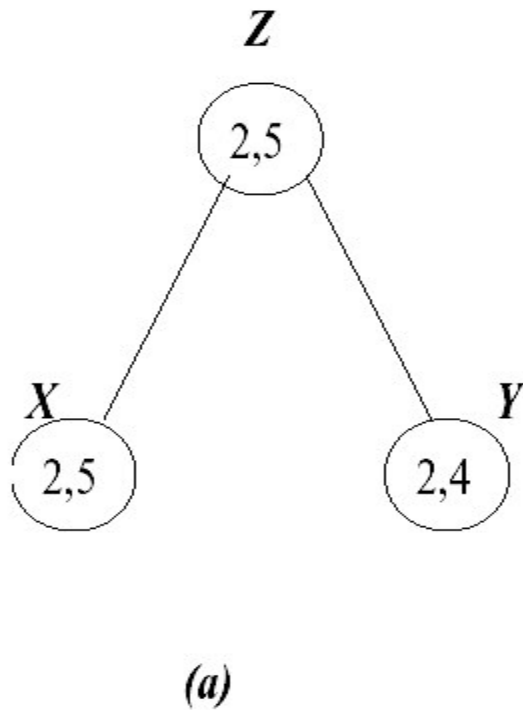
**output:**  $\mathcal{R}'$  which is the largest arc-consistent network equivalent to  $\mathcal{R}$

1. **for** every pair  $\{x_i, x_j\}$  that participates in a constraint  $R_{ij} \in \mathcal{R}$
2.      $queue \leftarrow queue \cup \{(x_i, x_j), (x_j, x_i)\}$
3. **endfor**
4. **while**  $queue \neq \{\}$
5.     select and delete  $(x_i, x_j)$  from  $queue$
6.      $Revise((x_i), x_j)$
7.     **if**  $Revise((x_i), x_j)$  causes a change in  $D_i$
8.         **then**  $queue \leftarrow queue \cup \{(x_k, x_i), i \neq k\}$
9.     **endif**
10. **endwhile**

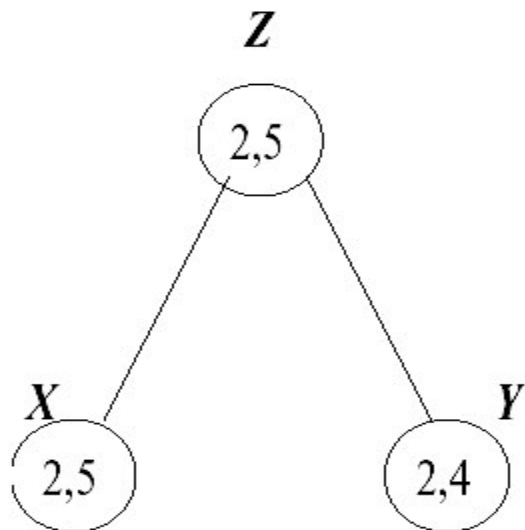
Figure 3.5: Arc-consistency-3 (AC-3)

- Complexity: since each arc may be processed in  $O(2k)$
- Best case  $O(ek)$ ,

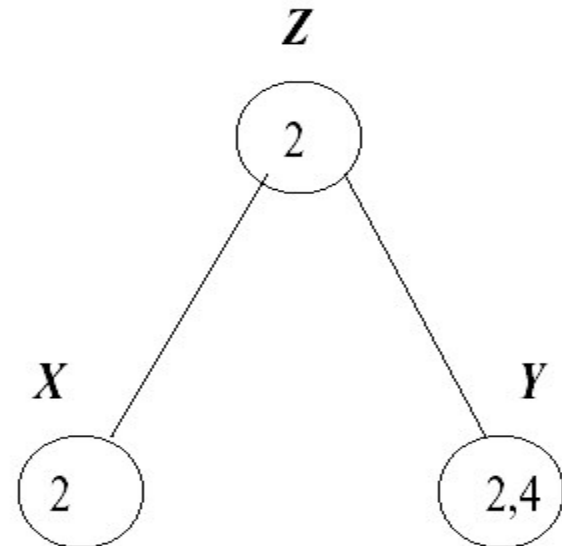
Example: a 3 variables network with 2 constraints:  $z$  divides  $x$  and  $z$  divides  $y$   
(a) before and (b) after AC-3 is applied.



Example: A 3 variables network with 2 constraints:  $z$  divides  $x$  and  $z$  divides  $y$   
(a) before and (b) after AC-3 is applied.



(a)



(b)

# AC-4

AC-4( $\mathcal{R}$ )

**input:** a network of constraints  $\mathcal{R}$

**output:** An arc-consistent network equivalent to  $\mathcal{R}$

1. Initialization:  $M \leftarrow \emptyset$ ,
2.     initialize  $S_{(x_i, a_i)}$ ,  $counter(i, a_i, j)$  for all  $R_{ij}$
3.     **for** all counters
4.         **if**  $counter(x_i, a_i, x_j) = 0$  (if  $\langle x_i, a_i \rangle$  is unsupported by  $x_j$ )
5.             **then** add  $\langle x_i, a_i \rangle$  to  $LIST$
6.         **endif**
7.     **endfor**
8. **while**  $LIST$  is not empty
9.     choose  $\langle x_i, a_i \rangle$  from  $LIST$ , remove it, and add it to  $M$
10.    **for** each  $\langle x_j, a_j \rangle$  in  $S_{(x_i, a_i)}$
11.       decrement  $counter(x_j, a_j, x_i)$
12.       **if**  $counter(x_j, a_j, x_i) = 0$
13.           **then** add  $\langle x_j, a_j \rangle$  to  $LIST$
14.       **endif**
15.    **endfor**
16. **endwhile**

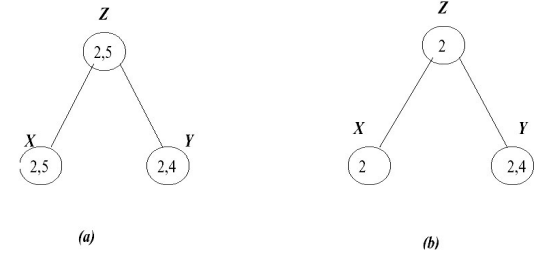


Figure 3.7: Arc-consistency-4 (AC-4)

- Complexity:
- (Counter is the number of supports to  $a_i$  in  $x_i$  from  $x_j$ .  $S_{(x_i, a_i)}$  is the set of pairs that  $(x_i, a_i)$  supports)

# Example applying AC-4

**Example 3.2.9** Consider the problem in Figure 3.6. Initializing the  $S_{(x,a)}$  arrays (indicating all the variable-value pairs that each  $\langle x, a \rangle$  supports), we have :

$$S_{(z,2)} = \{\langle x, 2 \rangle, \langle y, 2 \rangle, \langle y, 4 \rangle\}, S_{(z,5)} = \{\langle x, 5 \rangle\}, S_{(x,2)} = \{\langle z, 2 \rangle\}, \\ S_{(x,5)} = \{\langle z, 5 \rangle\}, S_{(y,2)} = \{\langle z, 2 \rangle\}, S_{(y,4)} = \{\langle z, 2 \rangle\}.$$

For counters we have:  $counter(x, 2, z) = 1$ ,  $counter(x, 5, z) = 1$ ,  $counter(z, 2, x) = 1$ ,  $counter(z, 5, x) = 1$ ,  $counter(z, 2, y) = 2$ ,  $counter(z, 5, y) = 0$ ,  $counter(y, 2, z) = 1$ ,  $counter(y, 4, z) = 1$ . (Note that we do not need to add counters between variables that are not directly constrained, such as  $x$  and  $y$ .) Finally,  $List = \{\langle z, 5 \rangle\}$ ,  $M = \emptyset$ . Once  $\langle z, 5 \rangle$  is removed from  $List$  and placed in  $M$ , the counter of  $\langle x, 5 \rangle$  is updated to  $counter(x, 5, z) = 0$ , and  $\langle x, 5 \rangle$  is placed in  $List$ . Then,  $\langle x, 5 \rangle$  is removed from  $List$  and placed in  $M$ . Since the only value it supports is  $\langle z, 5 \rangle$  and since  $\langle z, 5 \rangle$  is already in  $M$ , the  $List$  remains empty and the process stops.  $\square$

# Distributed arc-consistency (Constraint propagation)

- Implement AC-1 distributedly.
- $h_{j \rightarrow i}$  node  $x_j$  sends the message to node  $x_i$
- Node  $x_i$  updates its domain:
- Messages can be sent asynchronously or scheduled in a topological order



# Exercise: make the following network arc-consistent

- Draw the network's primal and dual constraint graph
- Network =
  - Domains  $\{1,2,3,4\}$
  - Constraints:  $y < x$ ,  $z < y$ ,  $t < z$ ,  $f < t$ ,  $x \leq t+1$ ,  $Y < f+2$

# Arc-consistency Algorithms

- **AC-1**: brute-force, distributed
- **AC-3**, queue-based
- **AC-4**, context-based, optimal
- **AC-5,6,7,....** Good in special cases
- **Important**: applied at every node of search
- ( $n$  number of variables,  $e$ =#constraints,  $k$ =domain size)
- Mackworth and Freuder (1977,1983), Mohr and Anderson, (1985)...

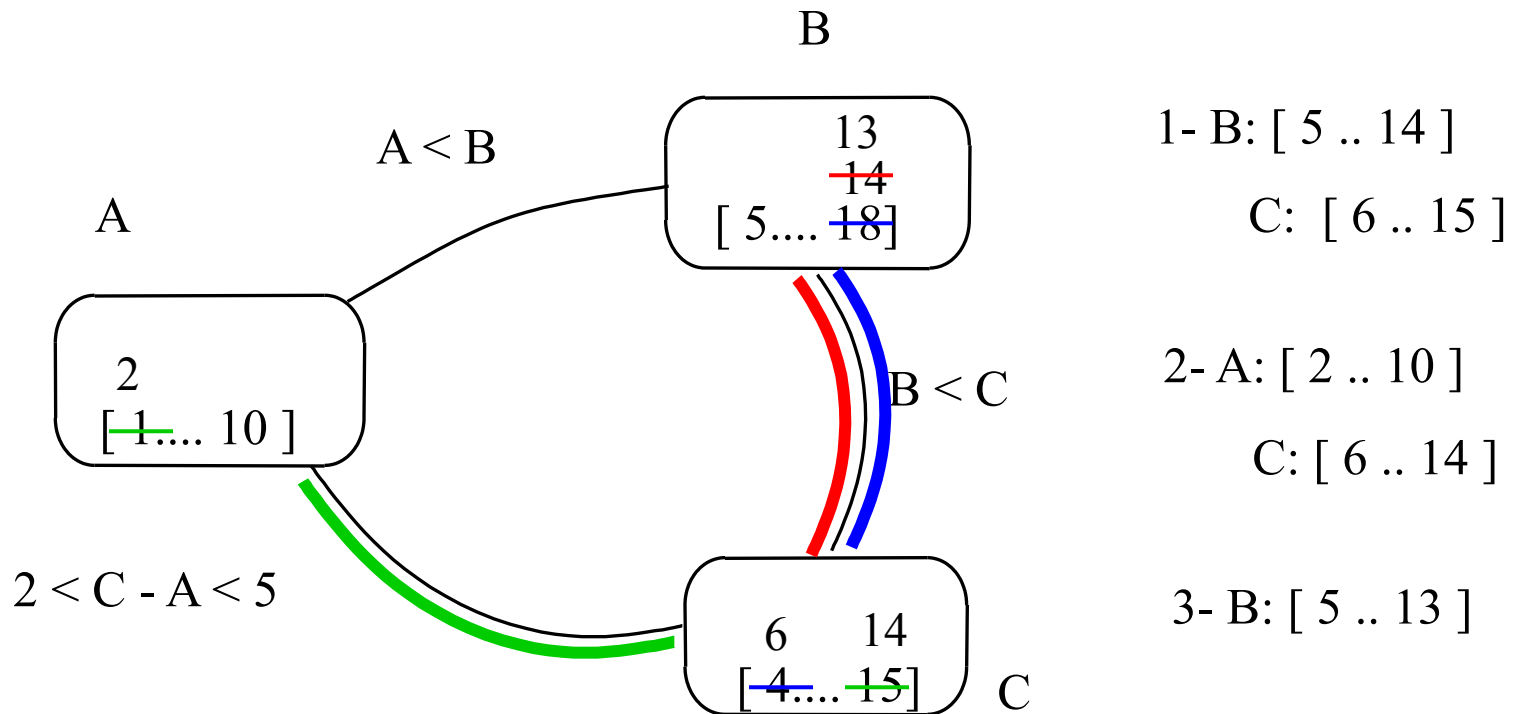
# Constraint tightness analysis

*t = number of tuples bounding a constraint*

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- **AC-3**, queue-based
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# Constraint checking

→ Arc-consistency



# Is arc-consistency enough?

- Example: a triangle graph-coloring with 2 values.
  - Is it arc-consistent?
  - Is it consistent?
- It is not path, or 3-consistent.

# Outline

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- Consistency operators: join, resolution, Gaussian elimination