

**Scribe Notes:** 2/13/2013  
**Presenter:** Dr. Berthe Y. Choueiry  
**Scribe:** Nate Stender  
**Topic:** Directional Consistency (Ch. 4 of Dechter book)

## Dechter's Slides<sup>1</sup> for Chapter 4 (Slides 31-37)

### Slide 31 - Adaptive-consistency, bucket-elimination

Adaptive-consistency algorithm

- Input: a constraint network  $R$  and an elimination ordering  $d$ .
- Output: a backtrack-free network  $E_d(R)$  along  $d$ , if the empty constraint was not generated. If empty constraint was generated, problem is inconsistent.

ADAPTIVE-CONSISTENCY (AC)

**Input:** a constraint network  $\mathcal{R}$ , an ordering  $d = (x_1, \dots, x_n)$

**output:** A backtrack-free network, denoted  $E_d(\mathcal{R})$ , along  $d$ , if the empty constraint was not generated. Else, the problem is inconsistent

1. Partition constraints into  $bucket_1, \dots, bucket_n$  as follows:  
     **for**  $i \leftarrow n$  **downto** 1, put in  $bucket_i$  all unplaced constraints mentioning  $x_i$ .
2. **for**  $p \leftarrow n$  **downto** 1 **do**
3.     **for** all the constraints  $R_{S_1}, \dots, R_{S_j}$  in  $bucket_p$  **do**
4.          $A \leftarrow \bigcup_{i=1}^j S_i - \{x_p\}$
5.          $R_A \leftarrow \Pi_A(\bigotimes_{i=1}^j R_{S_i})$
6.         **if**  $R_A$  is not the empty relation **then** add  $R_A$  to the bucket of the latest variable in scope  $A$ ,
7.         **else** exit and return the empty network
8. **return**  $E_d(\mathcal{R}) = (X, D, bucket_1 \cup bucket_2 \cup \dots \cup bucket_n)$

Figure 1: Dechter's adaptive-consistency algorithm

Comments:

- This algorithm is an example of dynamic programming:
- Problem is broken into sub-problems, and information is passed between the sub-problems
- Usually the goal of dynamic programming is to optimize a function
- We can apply this algorithm to Bayesian networks to find an optimal solution (the most probable explanation, MPE).
- Chapter 13 describes bucket elimination for solving optimization problems. Chapter 14 describes the application of bucket elimination for solving probabilistic networks

## Slide 32 - Adaptive-consistency, bucket-elimination (example)

Contrasting DPC, DiC and Adaptive Consistency (ADC)

- In DPC, given an ordering, generate binary constraints between all possible pairs of parents of the node (moralize the graph).
- In DiC, given an ordering, generate all possible constraints of arity  $(i-1)$  over the parents of the node.
- In ADC, given an ordering, we take all constraints between a node and its parents, join them, and project them over the parents, generating a new constraint involving all of the parents of (arity same as number of parents). **Guarantees tractability.**

*Tony and Daniel D. participated in a board example of bucket elimination*

Complexity:

- Time complexity is determined by the largest arity constraint generated:  $O(n(\exp)^{w^*(d)})$   
 $\Rightarrow O(n(k)^{w^*(d)+1})$
- Space complexity is determined by the largest arity constraint stored.
- Both complexities are improved by selecting an ordering with the smallest induced width (which will generate smaller maximum for arity of constraints).

## Slide 33 - Properties of bucket-elimination (adaptive consistency)

- Generates a **backtrack-free** constraint network.
- Time complexity:  $O(n(2k)^{w^*+1})$  (this is a refinement of time complexity mentioned before)
- Space complexity:  $O(n(k)^{w^*+1})$
- Special cases: **trees** ( $w^*=1$ ), **series-parallel networks** ( $w^*=2$ ), and in general **k-trees** ( $w^*=k$ ).

Partial answer to question by Daniel Dobos (<https://piazza.com/class#spring2013/csce921/20>) about k-trees:

*Question:* Section 4.1.3 covers *k-trees*, then the book never mentions them again. Is there anything else we should know about these, or was the section just included for completeness?

*Answer:* k-trees are mentioned because they can be used with ACD to provide a tractable solution.

## Slides 34 & 35 (moved over quickly because we have already covered them)

### Slide 36 - Summary: directional i-consistency

- Adaptive-consistency - creates constraint between all parents
- Directional path-consistency - creates a binary constraint between all pairs of parents
- Directional arc-consistency - filters domains of all parents

### Slide 37 - Variable Elimination

Full example of Bucket-elimination. Note that after one pass through the elimination order, the network can now be solved in a backtrack-free manner in the order. Now a solution can be created by assigning values to the variables in the instantiation ordering, which is the reverse of the elimination ordering.

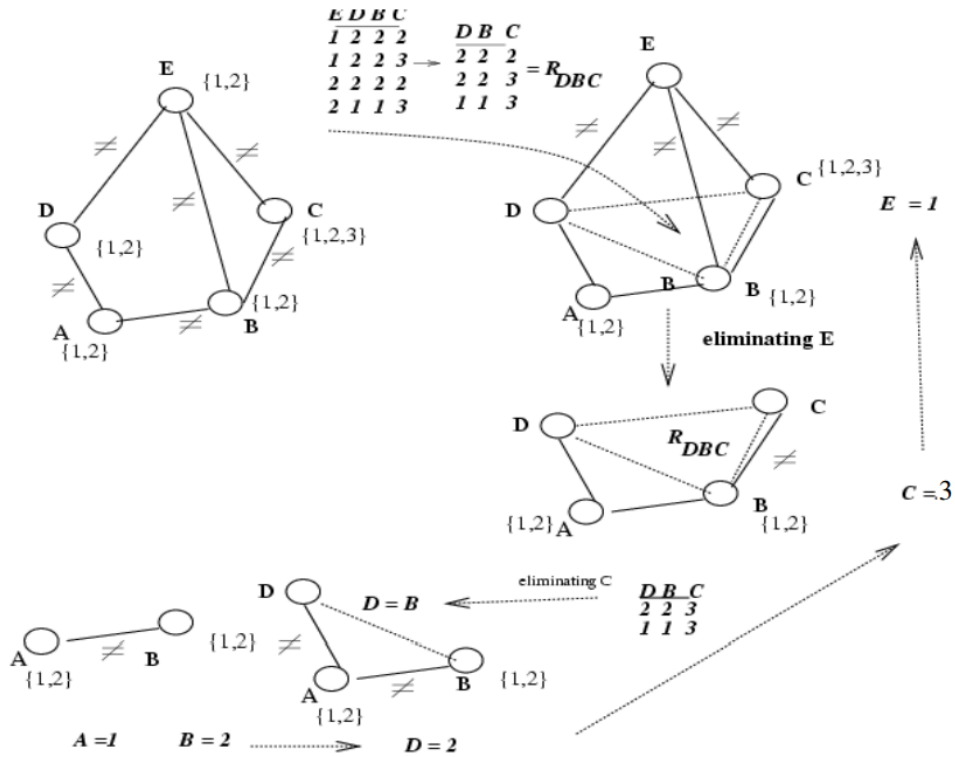


Figure 2: Full example of Bucket-elimination, followed by variable instantiation