

# Seminar: Search and Optimization

## Directional Consistency

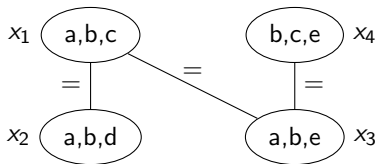
Gabi Röger

Universität Basel

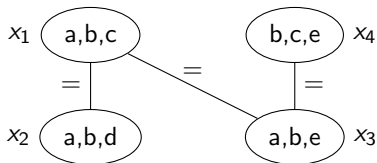
November 6, 2014

# Directional Arc-consistency

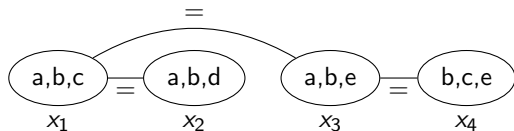
# Example



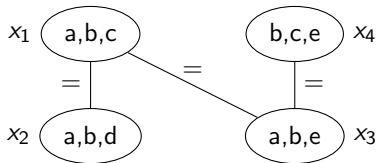
# Example



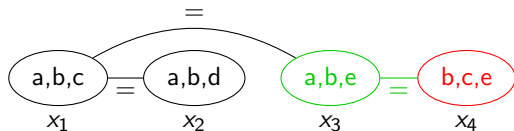
Assume we search with variable order  $x_1, x_2, x_3, x_4$



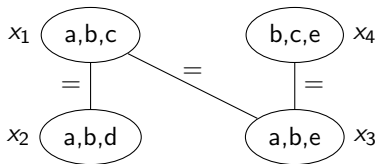
# Example



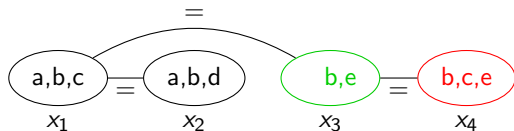
Assume we search with variable order  $x_1, x_2, x_3, x_4$



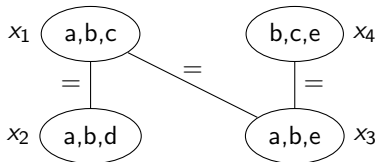
# Example



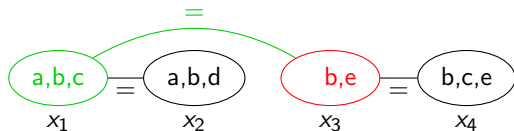
Assume we search with variable order  $x_1, x_2, x_3, x_4$



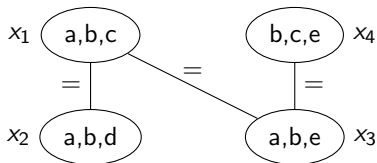
# Example



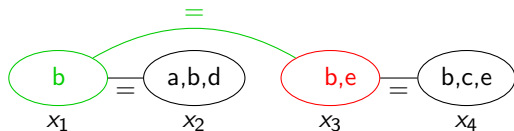
Assume we search with variable order  $x_1, x_2, x_3, x_4$



# Example

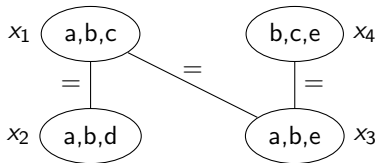


Assume we search with variable order  $x_1, x_2, x_3, x_4$

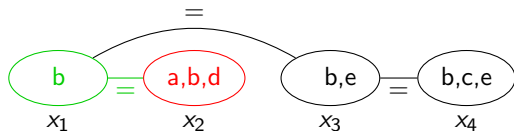




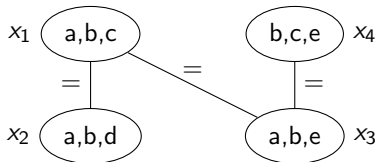
# Example



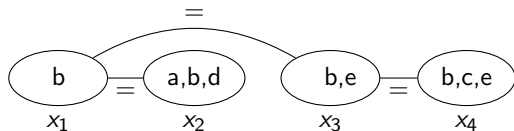
Assume we search with variable order  $x_1, x_2, x_3, x_4$



# Example

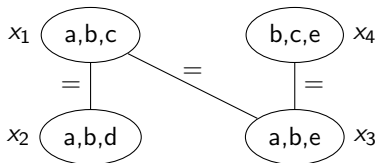


Assume we search with variable order  $x_1, x_2, x_3, x_4$

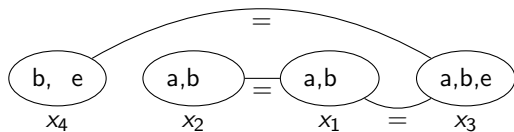


Backtrack-free search

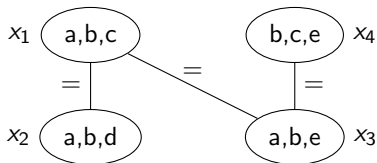
# Example



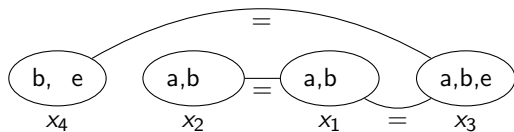
Assume we search with variable order  $x_4, x_2, x_1, x_3$



# Example



Assume we search with variable order  $x_4, x_2, x_1, x_3$



Not necessarily backtrack-free search

# Directional arc-consistency: Definition

## Definition (Directional arc-consistency)

A network is **directional arc-consistent** relative to variable order  $d = (x_1, \dots, x_n)$  iff every variable  $x_i$  is arc-consistent relative to every variable  $x_j$  such that  $i \leq j$ .

# Directional Arc-consistency: Function DAC

**function** DAC:

**for**  $i = n, \dots, 1$ :

**for each**  $j < i$  such that  $R_{ji} \in \mathcal{R}$ :

$D_j \leftarrow D_j \cap \pi_j(R_{ji} \bowtie D_i)$

        (remove values from  $D_j$  that don't have a partner in  $D_i$ )

**Input:** Constraint network  $\mathcal{R} = (X, D, C)$   
    with variable ordering  $d = (x_1, \dots, x_n)$

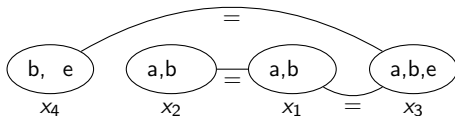
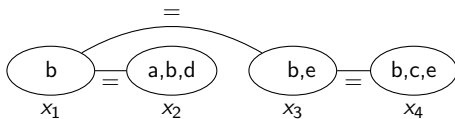
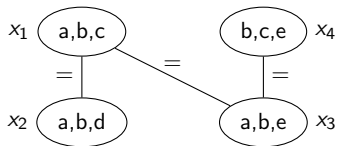
**Effect:** Enforces directional arc-consistency along  $d$ .

**Time complexity:**  $O(ek^2)$  with  $e$  binary constraints and  
    maximal domain size  $k$ .

## Directional Arc-consistency: Questions

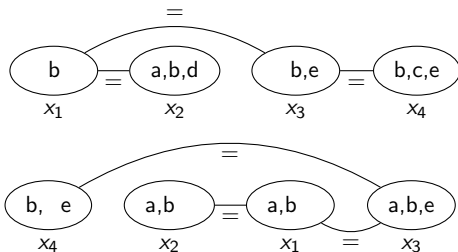
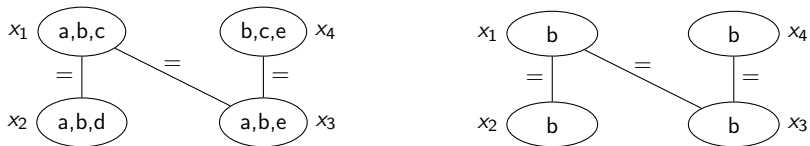
- How does directional arc-consistency relate to full arc-consistency?
- Is there a criterion when directional arc-consistency leads to backtrack-free search?
- Can we find a suitable variable ordering?

# Directional AC vs. Full AC

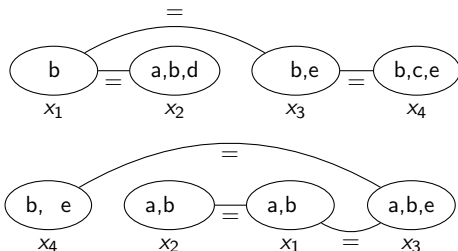
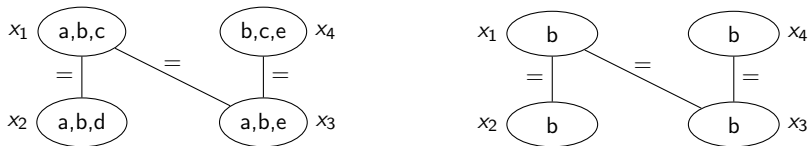




# Directional AC vs. Full AC



## Directional AC vs. Full AC



Enforcing full AC eliminates everything directional AC does  
... and more

# Width of a Graph

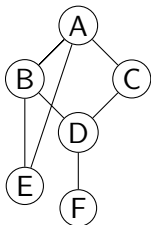
## Definition (Width of a graph)

Let  $G = (V, E)$  be an undirected graph and  $d = (v_1, \dots, v_n)$  be an ordering of its the nodes.

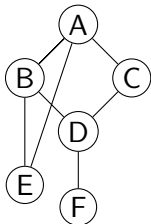
- The **parents** of a node  $v$  are its neighbours that precede it in the ordering.
- The **width of a node** is the number of its parents.
- The **width of the ordering** is the maximum width over all nodes.

The **width of graph  $G$**  is the minimum width over all orderings.

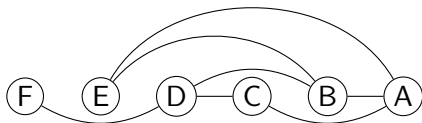
# Width of a graph: Example



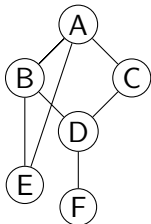
# Width of a graph: Example



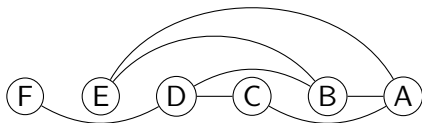
F,E,D,C,B,A:



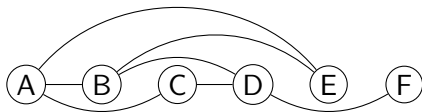
# Width of a graph: Example



F,E,D,C,B,A:



A,B,C,D,E,F:



# Width of Graph: Algorithm

**function** MIN-WIDTH:

$d \leftarrow$  array of size  $|V|$

**for**  $i = n, \dots, 1$ :

$r \leftarrow$  a node in  $G$  with smallest degree

$d[i] \leftarrow r$

    Remove all adjacent edges of  $r$  from  $E$

    Remove  $r$  from  $V$

**Input:** Graph  $G = (V, E)$

**Effect:**  $d$  contains minimum width ordering of nodes.

# Width of Graph and Directional Arc-Consistency

## Theorem

A graph is a *tree* iff it has *width 1*.

## Definition

A constraint network is *backtrack-free* relative to a given ordering  $(x_1, \dots, x_n)$  if for every  $i < n$ , every partial solutions of  $(x_1, \dots, x_i)$  can be consistently extended to include  $x_{i+1}$

## Theorem

Let  $d$  be a *width-1 ordering* of a constraint tree  $T$ . If  $T$  is *directional arc-consistent relative to  $d$*  then the network is *backtrack-free along  $d$* .



# Application: Algorithm for Trees

## function TREE-SOLVING:

Generate width-1 ordering  $(x_1, \dots, x_n)$  for  $\mathcal{R}$  along a rooted tree.

Let  $x_{p(i)}$  denote the parent of  $x_i$  in the rooted tree.

**for**  $i = n, \dots, 1$ :

$$D_{p(i)} \leftarrow D_{p(i)} \cap \pi_{p(i)}(R_{p(i)i} \bowtie D_i)$$

**if**  $D_{p(i)} = \emptyset$ :

    exit (inconsistent network)

Extract solution with (backtrack-free) search.

**Input:** Constraint network  $\mathcal{R} = (X, D, C)$

**Output:** Solution (or inconsistent network).

**Time complexity:**  $O(nk^2)$  with  $n$  variables and maximal domain size  $k$ .

# Directional Path-consistency

# (Strong) directional path-consistency: Function DPC

**function** DPC:

$E' \leftarrow E$

**for**  $k = n, \dots, 1$ :

**for each**  $i < k$  such that  $(x_i, x_k) \in E'$ :

$D_i \leftarrow D_i \cap \pi_i(R_{ik} \bowtie D_k)$

**for each**  $i, j < k$  such that  $(x_i, x_k), (x_j, x_k) \in E'$ :

$R_{ij} \leftarrow R_{ij} \cap \pi_{ij}(R_{ik} \bowtie D_k \bowtie R_{kj})$

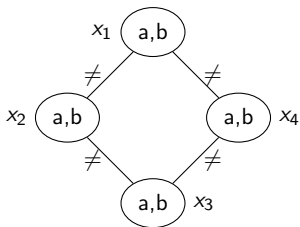
$E' \leftarrow E' \cup (x_i, x_j)$

**Input:** Constraint network  $\mathcal{R} = (X, D, C)$  with constraint graph  $G = (V, E)$  and variable ordering  $d = (x_1, \dots, x_n)$

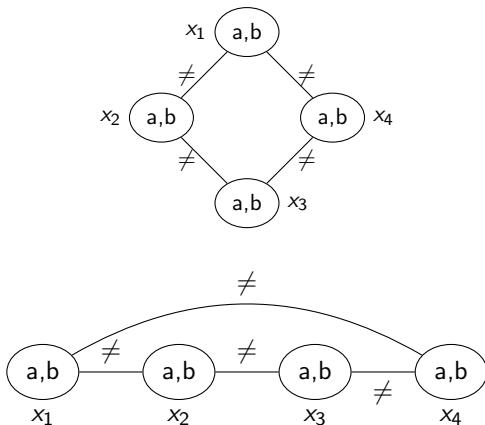
**Effect:** Enforces directional arc- and path-consistency along  $d$ .

**Time complexity:**  $O(n^3 k^3)$  with  $n$  variables and maximal domain size  $k$ .

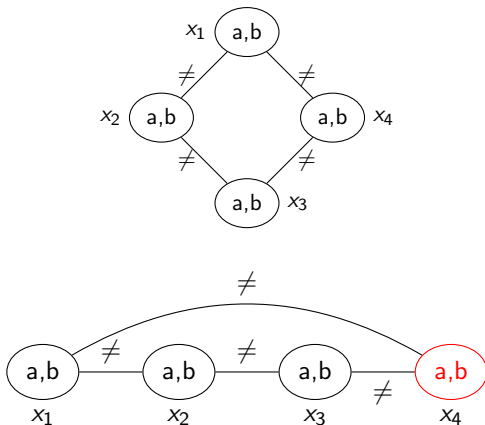
# Directional Path-consistency: Example



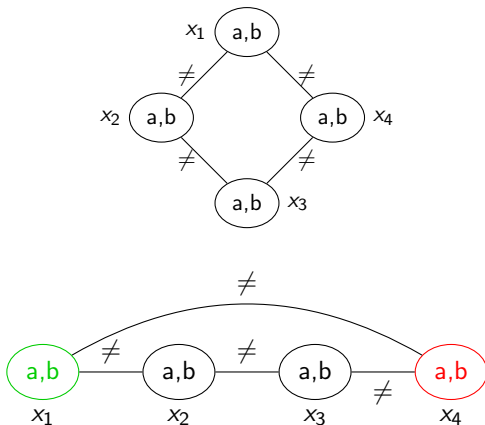
# Directional Path-consistency: Example



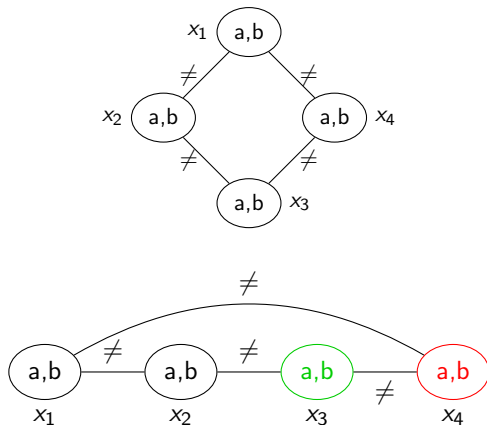
# Directional Path-consistency: Example



# Directional Path-consistency: Example

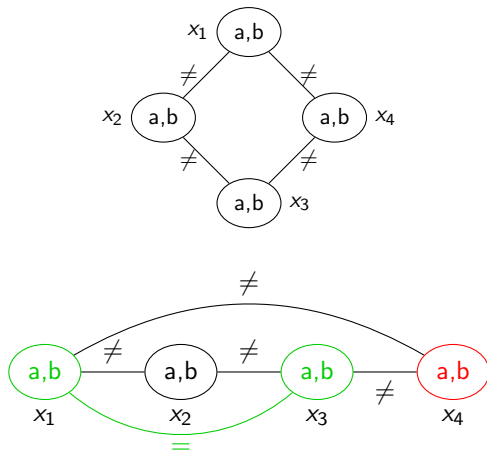


# Directional Path-consistency: Example

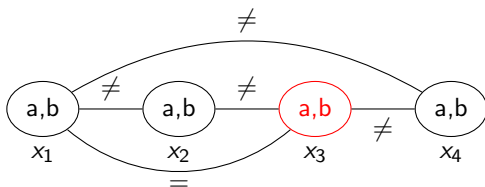
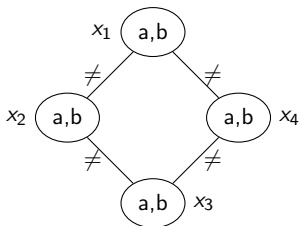




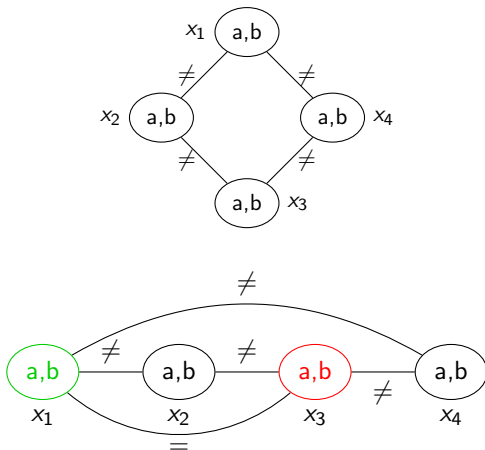
# Directional Path-consistency: Example



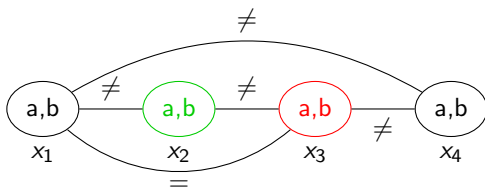
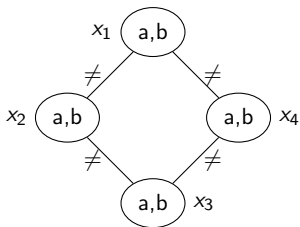
# Directional Path-consistency: Example



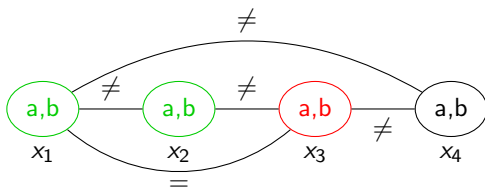
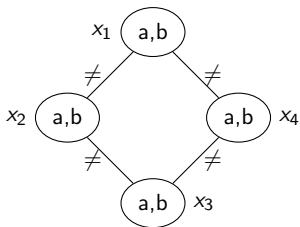
# Directional Path-consistency: Example



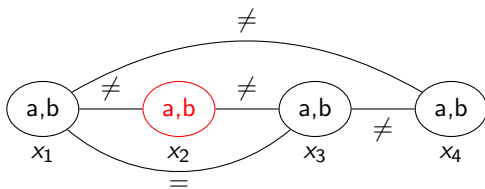
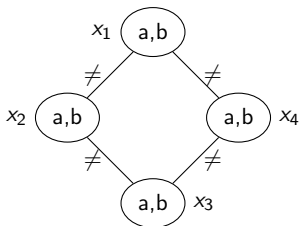
# Directional Path-consistency: Example



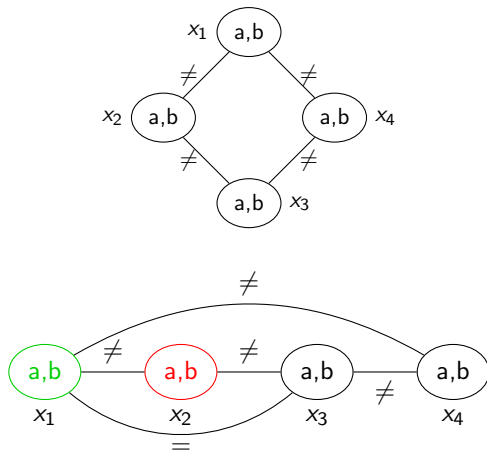
# Directional Path-consistency: Example



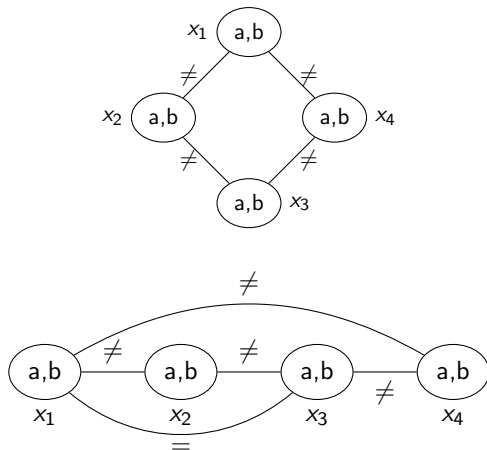
# Directional Path-consistency: Example



# Directional Path-consistency: Example



# Directional Path-consistency: Example





## Directional Path-consistency: Questions

- Is there a criterion when (strong) directional path-consistency leads to backtrack-free search?
- Can we find a suitable variable ordering?

## Directional Path-consistency: Questions

- Is there a criterion when (strong) directional path-consistency leads to backtrack-free search?
- Can we find a suitable variable ordering?
- ▶ Directional path-consistency can change the constraint graph.
- ▶ Width of constraint graph no longer sufficient.

# Directional Path-consistency: Questions

- Is there a criterion when (strong) directional path-consistency leads to backtrack-free search?
- Can we find a suitable variable ordering?
  - ▶ Directional path-consistency can change the constraint graph.
  - ▶ Width of constraint graph no longer sufficient.
  - ▶ Use **induced width** instead.

# Induced Width of a Graph

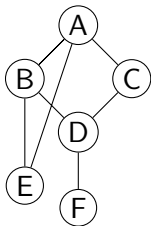
## Definition (Induced width of a graph)

Let  $G = (V, E)$  be an undirected graph and  $d = (v_1, \dots, v_n)$  be an ordering of its the nodes.

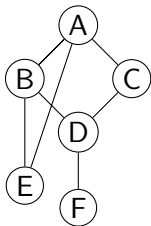
- Obtain graph  $G_d^*$  by processing the node ordering backwards and adding edges for each to parents of the processed node.
- The **induced width  $w_d^*$  of the ordering** is the width of  $G_d^*$ .

The **induced width  $w^*$  of graph  $G$**  is the minimal induced width over all orderings.

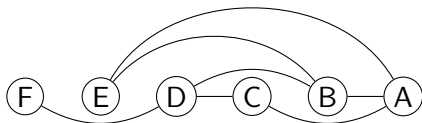
# Induced Width of a Graph: Example



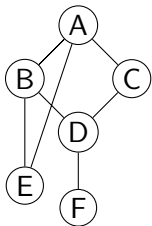
# Induced Width of a Graph: Example



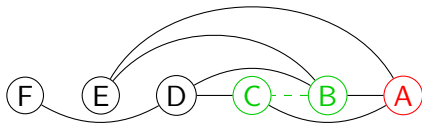
F,E,D,C,B,A:



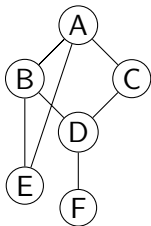
# Induced Width of a Graph: Example



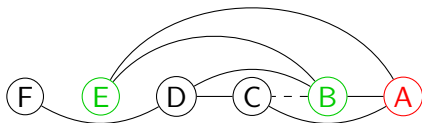
F, E, D, C, B, A:



# Induced Width of a Graph: Example

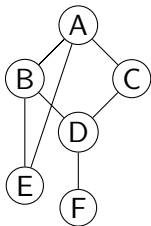


F, E, D, C, B, A:

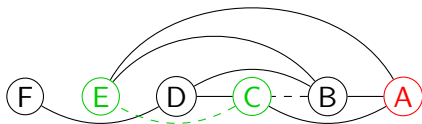




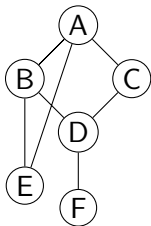
# Induced Width of a Graph: Example



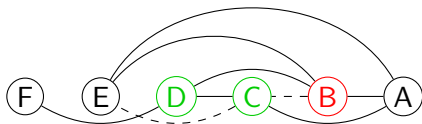
F, E, D, C, B, A:



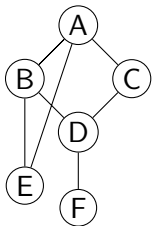
# Induced Width of a Graph: Example



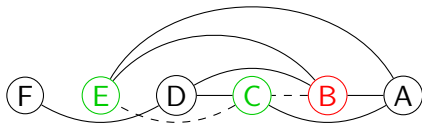
F, E, D, C, B, A:



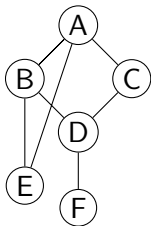
# Induced Width of a Graph: Example



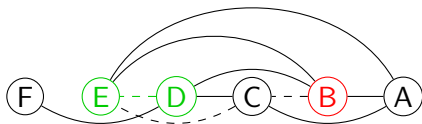
F, E, D, C, B, A:



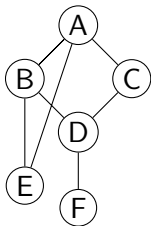
# Induced Width of a Graph: Example



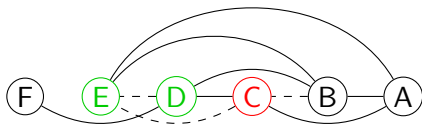
F, E, D, C, B, A:



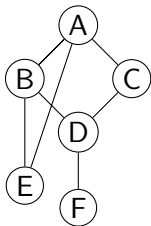
# Induced Width of a Graph: Example



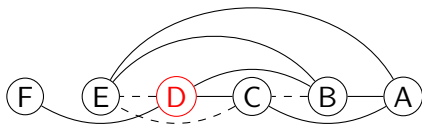
F, E, D, C, B, A:



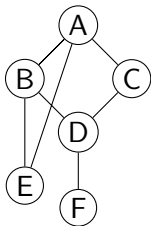
# Induced Width of a Graph: Example



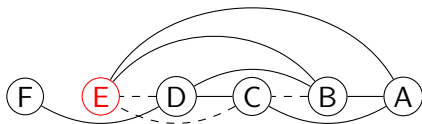
F, E, D, C, B, A:



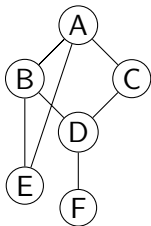
# Induced Width of a Graph: Example



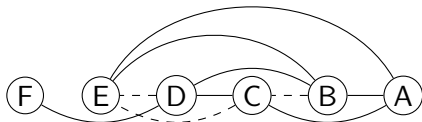
F, E, D, C, B, A:



# Induced Width of a Graph: Example



F, E, D, C, B, A:



Induced width  $w_{(F,E,D,C,B,A)}^*$ : 3



# Induced Width of Graph: Algorithm 1

- Determining the induced width of a graph is **NP-hard**
- Find **good** ordering in polynomial time

# Induced Width of Graph: Algorithm 1

- Determining the induced width of a graph is **NP-hard**
- Find **good** ordering in polynomial time

## function MIN-DEGREE:

$d \leftarrow$  array of size  $|V|$

**for**  $i = n, \dots, 1$ :

$r \leftarrow$  a node in  $G$  with smallest degree

$d[i] \leftarrow r$

**Connect  $r$ 's parents:**  $E \leftarrow E \cup \{(v, v') \mid (v, r), (v', r) \in E\}$

    Remove all adjacent edges of  $r$  from  $E$

    Remove  $r$  from  $V$

**Input:** Graph  $G = (V, E)$

**Effect:**  $d$  contains ordering with small induced width.

# Induced Width of Graph: Algorithm 2

**function** MIN-FILL: $d \leftarrow$  array of size  $|V|$ **for**  $i = n, \dots, 1$ : $r \leftarrow$  a node in  $G$  with **fewest missing edges between parents** $d[i] \leftarrow r$ Connect  $r$ 's parents:  $E \leftarrow E \cup \{(v, v') \mid (v, r), (v', r) \in E\}$ Remove all adjacent edges of  $r$  from  $E$ Remove  $r$  from  $V$ **Input:** Graph  $G = (V, E)$ **Effect:**  $d$  contains ordering with small induced width.

# Width of Graph and Directional Arc-Consistency

## Theorem

*Let  $G$  be the constraint graph of a binary network  $\mathcal{R}$  and let  $d$  be a variable ordering. If DPC is applied to  $\mathcal{R}$  with ordering  $d$  then the resulting constraint graph is subsumed by the Graph  $G_d^*$ .*

# Width of Graph and Directional Arc-Consistency

## Theorem

*Let  $G$  be the constraint graph of a binary network  $\mathcal{R}$  and let  $d$  be a variable ordering. If DPC is applied to  $\mathcal{R}$  with ordering  $d$  then the resulting constraint graph is subsumed by the Graph  $G_d^*$ .*

## Theorem

*Given a binary network  $\mathcal{R}$  and an ordering  $d$ , the time complexity of DPC along  $d$  is  $O((w_d^*)^2 \cdot n \cdot k^3)$ .*

# Width of Graph and Directional Arc-Consistency

## Theorem

*Let  $G$  be the constraint graph of a binary network  $\mathcal{R}$  and let  $d$  be a variable ordering. If DPC is applied to  $\mathcal{R}$  with ordering  $d$  then the resulting constraint graph is subsumed by the Graph  $G_d^*$ .*

## Theorem

*Given a binary network  $\mathcal{R}$  and an ordering  $d$ , the time complexity of DPC along  $d$  is  $O((w_d^*)^2 \cdot n \cdot k^3)$ .*

Previously:  $O(n^3 k^3)$

Lesson learned: Prefer orderings with small induced width

# Adaptive Consistency

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional *i*-consistency**.



# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .
- Algorithm idea for CSP solving:

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .
- Algorithm idea for CSP solving:
  - 1 Select ordering  $d$  with small width.

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .
- Algorithm idea for CSP solving:
  - 1 Select ordering  $d$  with small width.
  - 2 Compute its induced width  $w_d^*$ .

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .
- Algorithm idea for CSP solving:
  - 1 Select ordering  $d$  with small width.
  - 2 Compute its induced width  $w_d^*$ .
  - 3 Apply strong directional  $w_d^* + 1$ -consistency.

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .
- Algorithm idea for CSP solving:
  - 1 Select ordering  $d$  with small width.
  - 2 Compute its induced width  $w_d^*$ .
  - 3 Apply strong directional  $w_d^* + 1$ -consistency.
  - 4 Determine solution with backtrack-free search.

# Motivation

- Concept of directional arc- and path-consistency can be generalized to **directional  $i$ -consistency**.
- If a network  $\mathcal{R}$  has **induced width  $i - 1$  for ordering  $d$**  and it is **strong directional  $i$ -consistent for  $d$**  then  $\mathcal{R}$  is **backtrack-free along  $d$** .
- Algorithm idea for CSP solving:
  - 1 Select ordering  $d$  with small width.
  - 2 Compute its induced width  $w_d^*$ .
  - 3 Apply strong directional  $w_d^* + 1$ -consistency.
  - 4 Determine solution with backtrack-free search.
- Idea: Combine steps 2 and 3

# Adaptive Consistency: Function ADC

## function ADC:

$E' \leftarrow E, C' \leftarrow C$

**for**  $k = n, \dots, 1$ :

$S \leftarrow$  parents of  $x_k$  w.r.t.  $E'$  and  $d$

$R_S \leftarrow \text{REVISE}(S, x_k)$

$C' \leftarrow C' \cup R_S$

$E' \leftarrow E' \cup \{(x_i, x_j) \mid x_i, x_j \in S, x_i \neq x_j\}$

**Input:** Constraint network  $\mathcal{R} = (X, D, C)$  with constraint graph  $G = (V, E)$  and variable ordering  $d = (x_1, \dots, x_n)$

**Effect:** Enforces strong directional  $w_d^* + 1$ -consistency and the resulting network has width bounded by  $w_d^*$ .  
 $\mathcal{R}$  consistent  $\Rightarrow$  resulting network backtrack-free along  $d$ .

**Time complexity:**  $O(n \cdot (2k)^{w_d^*+1})$



# Tractable Class of Constraint Satisfaction Problems

If the induced width for a problem is bounded by a constant  $b$ , we can efficiently find an ordering  $d$  with  $w_d^* \leq b$ .

# Tractable Class of Constraint Satisfaction Problems

If the induced width for a problem is bounded by a constant  $b$ , we can efficiently find an ordering  $d$  with  $w_d^* \leq b$ .

## Theorem

*The class of constraint problems whose induced width is bounded by a constant  $b$  is solvable in polynomial time and space.*

# Summary

# Summary

- **Directional arc- and path-consistency** can be used as preprocessing algorithm or for interleaved reasoning during search.
- Guarantee **backtrack-free** search for problems with induced width 1 (for directional arc-consistency) and 2 (for strong directional path-consistency), respectively.
- Identified a **tractable class of constraint satisfaction problems**
- Purely **structural** criterion: induced width of constraint graph