



# An algebraic notion of conditional independence, and its application to knowledge representation

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## Motivating Example

$r : \text{inf}(X) \leftarrow \text{inf}(Y), \text{cnct}(Y, X), \text{not vac}(X).$

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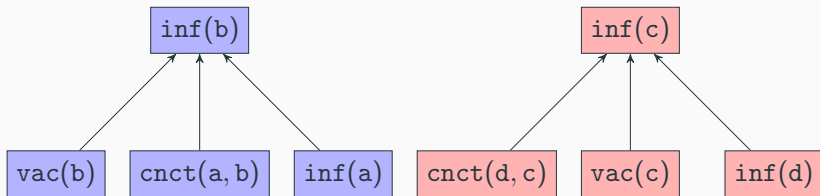
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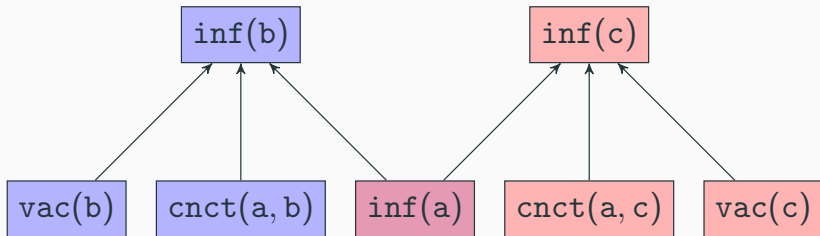
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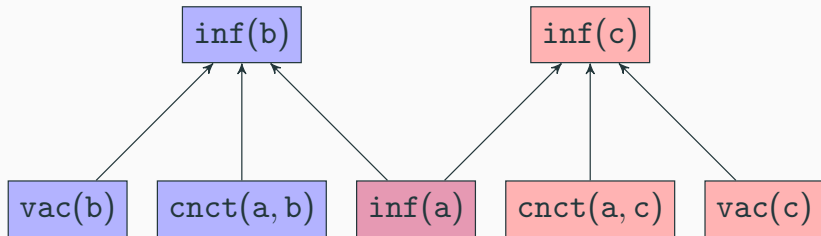


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Once we know  $\text{inf}(a)$  (or  $\neg \text{inf}(a)$ ), we obtain two independent subprograms.

- Should we investigate conditional independence for normal logic programs? Or logic programs with aggregates, disjunction, choice constructs, ...



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- Should we investigate conditional independence for normal logic programs? Or logic programs with aggregates, disjunction, choice constructs, ...
- Under the stable semantics? Or the well-founded, partial stable, supported, regular, ... model semantics?
- And what about abstract (dialectical) argumentation (with weights?), autoepistemic logic, default logic, ...?
- Conditional independence seems a good candidate for a **top-down** approach.

- Constructive techniques for approximating the fixpoints of an operator  $O$  over a lattice  $L$ .
- Uniform framework for the mechanisms underlying many different knowledge representation formalisms, such as logic programming [PDB07], autoepistemic logic [DMT03], default logic [DMT03], abstract argumentation [SW15] and abstract dialectical frameworks [SW15].
- To define semantics for a formalism, the user merely has to choose the lattice and define the operator, and then AFT does all the hard work for the user.

# Lattices and operators

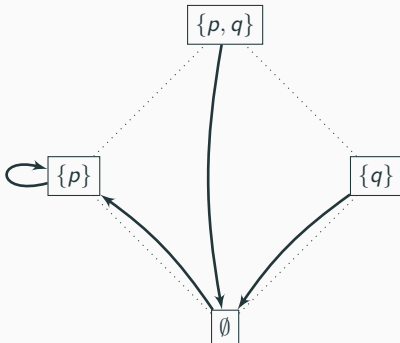
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$$\mathcal{P} = \{p \leftarrow \neg q\}$$

$$\text{IC}_{\mathcal{P}}(x) = \{\alpha \in \mathcal{A}_{\mathcal{P}} \mid \alpha \leftarrow \phi \in \mathcal{P} \text{ and } x(\phi) = \text{T}\}$$

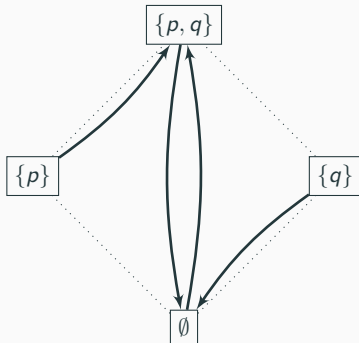


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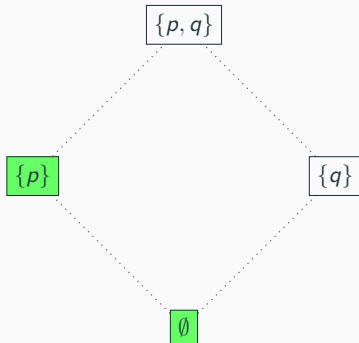


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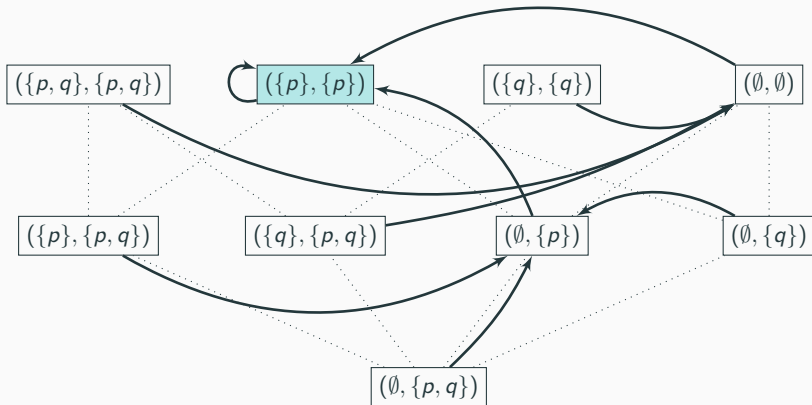
$$\mathcal{IC}_{\mathcal{P}}(x, y) = \{\alpha \in \mathcal{A}_{\mathcal{P}} \mid \alpha \leftarrow \phi \in \mathcal{P} \text{ and } (x, y)(\phi) \geq \top\}$$





# Lattices and operators

Given a lattice  $\mathcal{L} = \langle L, \leq \rangle$ , we are interested in operator  $O : L \rightarrow L$  and its fixpoints.



## AFT: summary

- Represent or define NMR-formalism by specifying a **lattice** and a (family of) **operator**(s).
- Many operators are **non-monotonic** and obtain a natural semantics in terms of **approximations**, i.e. pairs of elements.
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Formalism	Lattice	Operator
Logic programming	Herbrand bases ordered by $\subseteq$	Immediate consequence
Formal argumentation	Sets of arguments ordered by $\subseteq$	Defense
ADFs	Sets of arguments ordered by $\subseteq$	$\Gamma$ -operator
Default logic	Sets of possible worlds ordered by $\supseteq$	Immediate consequence
Weighted ADFs	Multivalued assignments	$\Gamma$ -operator
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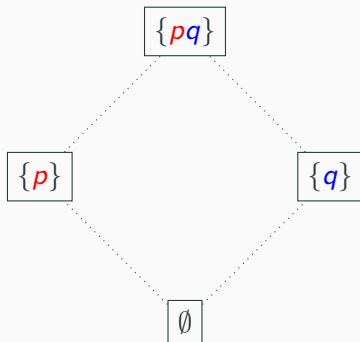
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If we study a concept in AFT, it is applicable to many formalisms under a wide family of semantics.

# Roadmap

We are interested in defining conditional independence of *parts of a lattice* sanctioned by an operator.

1. Divide lattices in sub-lattices
2. Define and study conditional independence of sub-lattices w.r.t.  $O$ .



## Sub-lattices

Let  $I$  be a set, which we call the *index set*, and for each  $i \in I$ , let  $L_i$  be a set. The product set  $\bigotimes_{i \in I} L_i$  is the following set of functions:

$$\bigotimes_{i \in I} L_i = \{f \mid f : I \rightarrow \bigcup_{i \in I} L_i \text{ s.t. } \forall i \in I : f(i) \in L_i\}$$

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The product set  $\bigotimes_{i \in I} L_i$  contains all ways of selecting one element of every set  $L_i$ .

For a finite set  $I = \{1, \dots, n\}$ , the product  $\bigotimes_{i \in I} L_i$  is (isomorphic to) the cartesian product  $L_1 \times \dots \times L_n$ .

### Example

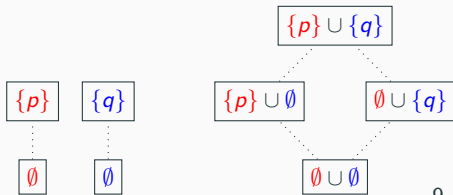
$L_1 = \{\emptyset, \{p\}\}$  and  $L_2 = \{\emptyset, \{q\}\}$ ,

$\bigotimes_{i \in \{1,2\}} L_i$  contains

$f(1) = f(2) = \emptyset$ , and

$f'(1) = \emptyset$  and  $f'(2) = \{q\}$ ,

...



- Motivation and Preliminaries
- Sub-lattices
- Conditional Independence w.r.t. an operator
- Results
- Application to Logic Programming



# Conditional Independence w.r.t. an operator

## Definition

Let  $O : L_1 \otimes L_2 \otimes L_3 \rightarrow L_1 \otimes L_2 \otimes L_3$  be given.

$L_1 \perp\!\!\!\perp_O L_2 \mid L_3$  if there exist operators

$$O_{1,3} : L_1 \otimes L_3 \rightarrow L_1 \otimes L_3 \text{ and } O_{2,3} : L_2 \otimes L_3 \rightarrow L_2 \otimes L_3$$

s.t. for  $i, j \in \{1, 2\}$ ,  $i \neq j$ , and for every  $x_i \otimes x_3 \in L_i \otimes L_3$  and for every  $x_j \in L_j$  it holds that:

$$O(x_i \otimes x_j \otimes x_3)_{|i,3} = O_{i,3}(x_i \otimes x_3).$$

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where  $x_i \otimes x_j \otimes x_3|_{i,3} = x_i \otimes x_3$

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$\mathcal{A}_1 = \{\text{inf}(\text{b}), \text{cnct}(\text{a}, \text{b}), \text{vac}(\text{b})\}$

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$$\mathcal{P}_1 = \{r_1, r_3, r_4\} \text{ and } \mathcal{P}_2 = \{r_2, r_3, r_5\}$$

For any  $x_1 \subseteq \mathcal{A}_1$ ,  $x_2 \subseteq \mathcal{A}_2$ ,  $x_3 \subseteq \mathcal{A}_3$ ,

$$IC_{\mathcal{P}}(x_1 \cup x_2 \cup x_3) \cap (\mathcal{A}_1 \cup \mathcal{A}_3) = IC_{\mathcal{P}_1}(x_1 \cup x_2)$$

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(similarly for  $\mathcal{P}_2$ ), and thus:

$$2^{\mathcal{A}_1} \perp\!\!\!\perp_{IC_{\mathcal{P}}} 2^{\mathcal{A}_2} \mid 2^{\mathcal{A}_3}$$

$$A \perp\!\!\!\perp B|C \text{ if } P(A, B|C) = P(A|C)P(B|C).$$

## Similarities with Probability Theory

$$A \perp\!\!\!\perp B|C \text{ if } P(A, B|C) = P(A|C)P(B|C).$$

### Fact

$L_1 \perp\!\!\!\perp_O L_2 \mid L_3$  implies:

$$\begin{aligned} O(x_1 \otimes x_2 \otimes x_3) &= O_{1,3}(x_1 \otimes x_3) \otimes O_{2,3}(x_2 \otimes x_3)|_2 \\ &= O_{1,2}(x_1 \otimes x_3)|_1 \otimes O_{2,3}(x_2 \otimes x_3) \end{aligned}$$



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Furthermore, for any  $i, j = 1, 2, i \neq j, x_i \in \mathcal{L}_i, x_j, x'_j \in \mathcal{L}_j$  and  $x_3 \in \mathcal{L}_3$  it holds that:

$$O(x_i \otimes x_j \otimes x_3)|_{i,3} = O(x_i \otimes x'_j \otimes x_3)|_{i,3}$$

## Proposition

Let an operator  $O$  s.t.  $L_1 \perp\!\!\!\perp_O L_2 \mid L_3$  be given.

Then  $x_1 \otimes x_2 \otimes x_3 = O(x_1 \otimes x_2 \otimes x_3)$  iff

$x_1 \otimes x_3 = O_{1,3}(x_1 \otimes x_3)$  and  $x_2 \otimes x_3 = O_{2,3}(x_2 \otimes x_3)$ .

# Search for Fixpoints can be Split

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We can look for supported models of  $\mathcal{P}$  by looking for supported models of  $\mathcal{P}_1$  and  $\mathcal{P}_2$  and combining them afterwards.

# Monotonicity is Preserved

## Proposition

Let an operator  $O$  s.t.  $L_1 \perp\!\!\!\perp_O L_2 \mid L_3$  be given.

Then  $O : \bigotimes_{i \in \{1,2,3\}} L_i \rightarrow \bigotimes_{i \in \{1,2,3\}} L_i$  is  $\leq_{\otimes}$ -monotonic iff

$O_{i,3} : L_i \otimes L_3 \rightarrow L_i \otimes L_3$  is  $\leq_{\otimes}^{i,3}$ -monotonic for  $i = 1, 2$ .

## Proposition

Let a  $\leq_{\otimes}$ -monotonic operator  $O$  s.t.  $L_1 \perp\!\!\!\perp_O L_2 \mid L_3$  be given.

Then  $x$  is a least fixed point of  $O$  iff  $x_{|i,3}$  is a least fixed point of

$O_{i,3}$  (for  $i = 1, 2$ ).

## Proposition

Let  $\mathcal{O}$  be an approximation operator s.t.  $L_1^2 \perp\!\!\!\perp_{\mathcal{O}} L_2^2 \mid L_3^2$ . Then:

- $(x, y)$  is the Kripke-Kleene fixpoint of  $\mathcal{O}$  iff  $(x_{|i,3}, y_{|i,3})$  is the Kripke-Kleene fp. of  $\mathcal{O}_{i,3}$  for  $i = 1, 2$ .
- $(x, y)$  is a fixpoint of  $\mathcal{O}$  iff  $(x_{|i,3}, y_{|i,3})$  is a fp. of  $\mathcal{O}_{i,3}$  for  $i = 1, 2$ .

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$L_1^2 \perp\!\!\!\perp_{\mathcal{C}(\mathcal{O}_l)} L_2^2 \mid L_3^2$  and  $L_1^2 \perp\!\!\!\perp_{\mathcal{C}(\mathcal{O}_u)} L_2^2 \mid L_3^2$ .

### Proposition

Let  $\mathcal{O}$  be an approximation operator s.t.  $L_1^2 \perp_{\mathcal{O}} L_2^2 \mid L_3^2$ . Then:

1.  $(x, y)$  is a fixpoint of  $S(\mathcal{O})$  iff  $(x_{|i,3}, y_{|i,3})$  is a fp. of  $S(\mathcal{O}_{i,3})$  for  $i = 1, 2$ .
2.  $(x, y)$  is the well-founded fixpoint of  $\mathcal{O}$  iff  $(x_{|i,3}, y_{|i,3})$  is the well-founded fp. of  $\mathcal{O}_{i,3}$  for  $i = 1, 2$ .

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These results show that most reasoning tasks in NMR can be tackled using a *divide-and-conquer*-methodology if the problem instance admits conditional independencies.

## Proposition

Let a normal logic program  $\mathcal{P}$  be given for which  $\mathcal{A}_{\mathcal{P}}$  is partitioned into  $\mathcal{A}_1 \cup \mathcal{A}_2 \cup \mathcal{A}_3$  s.t.  $\mathcal{A}_1 \perp\!\!\!\perp_{\mathcal{P}} \mathcal{A}_2 \mid \mathcal{A}_3$ .

- $(x_1 \cup x_2 \cup x_3, y_1 \cup y_2 \cup y_3)$  is a supported model of  $\mathcal{P}$  iff  $(x_i \cup x_3, y_i \cup y_3)$  is a supported model of  $\mathcal{P}_{|\mathcal{A}_i \cup \mathcal{A}_3}$  (for  $i = 1, 2$ ).
- $(x_1 \cup x_2 \cup x_3, y_1 \cup y_2 \cup y_3)$  is a 3-valued stable model of  $\mathcal{P}$  iff  $(x_i \cup x_3, y_i \cup y_3)$  is a 3-valued stable model of  $\mathcal{P}_{|\mathcal{A}_i \cup \mathcal{A}_3}$  (for  $i = 1, 2$ ).
- The [ultimate] well-founded model of  $\mathcal{P}$  can be obtained as  $(x_1 \cup x_2 \cup x_3, y_1 \cup y_2 \cup y_3)$ , where  $(x_i \cup x_3, y_i \cup y_3)$  is the well-founded model of  $\mathcal{P}_{\mathcal{A}_j}$  (for  $i, j = 1, 2, i \neq j$ ).



# Splitting up Reasoning Tasks

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## Example

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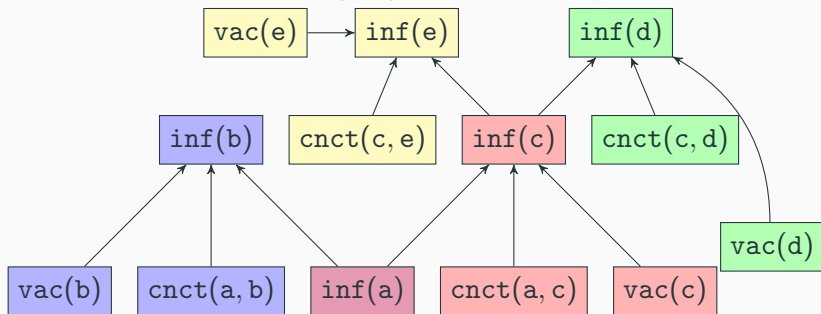
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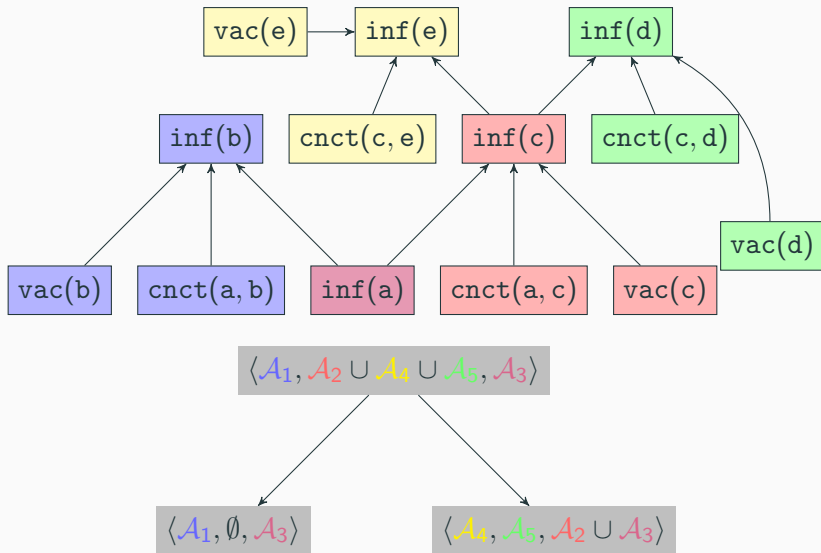
$r_6$  :  $\text{inf}(d) \leftarrow \text{inf}(c), \text{cnct}(c, d), \text{not vac}(d).$

$r_7$  :  $\text{inf}(e) \leftarrow \text{inf}(c), \text{cnct}(c, e), \text{not vac}(e).$

$r_8$  :  $\text{cnct}(c, d).$ ,  $r_9$  :  $\text{cnct}(c, e).$

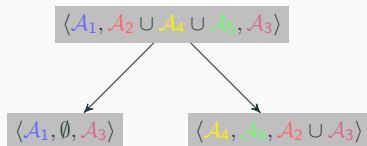


# Example

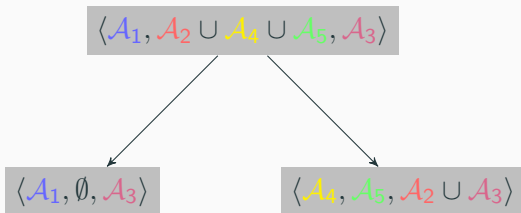


A binary labelled tree  $(V, E, \nu)$  is a *conditional independence tree* for  $\mathcal{O}$  (in short, CIT) if the following holds:

- $\nu : V \rightarrow 2^I \times 2^I \times 2^I$ ,
- the root is labelled  $\langle l_1, l_2, l_3 \rangle$  where  $l_1, l_2, l_3$  is a partition of  $I$ ,
- for every  $(v_1, v_2), (v_1, v_3) \in E$ , where  $\nu(v_i) = \langle l_1^i, l_2^i, l_3^i \rangle$  for  $i = 1, 2, 3$ ,  
 $l_j^1 \cup l_3^1 = l_1^j \cup l_j^2 \cup l_j^3$  for  $j = 2, 3$  and  
 $l_j^1 \cup l_2^1 \cup l_3^1 = l_1^2 \cup l_2^2 \cup l_3^2 \cup l_1^3 \cup l_2^3 \cup l_3^3$ ,
- if  $\nu(v) = \langle l_1, l_2, l_3 \rangle$  then  $\bigotimes_{i \in l_1} \mathcal{L}_i \perp\!\!\!\perp_{\mathcal{O}_{l_1 \cup l_2 \cup l_3}} \bigotimes_{i \in l_2} \mathcal{L}_i \mid \bigotimes_{i \in l_3} \mathcal{L}_i$ .

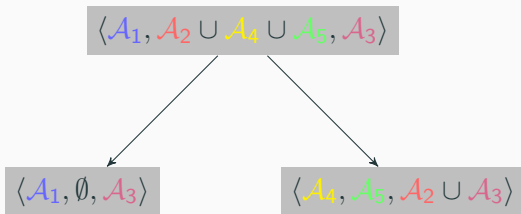


## Example: Well-Founded Interpretation



- $\text{WF}(\mathcal{P}_1 \cup \mathcal{P}_3) = \{\text{inf}(a), \text{cnct}(a, b), \text{inf}(b)\}$   
Search space size:  $2^4$
- $\text{WF}(\mathcal{P}_4 \cup \mathcal{P}_2 \cup \mathcal{P}_3) = \{\text{inf}(a), \text{cnct}(a, c), \text{inf}(c), \text{cnct}(c, e), \text{inf}(e)\}$   
Search space size:  $2^7$
- $\text{WF}(\mathcal{P}_5 \cup \mathcal{P}_2 \cup \mathcal{P}_3) = \{\text{inf}(a), \text{cnct}(a, c), \text{inf}(c), \text{cnct}(c, d), \text{inf}(d)\}$   
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Search space size:  $2^7$
- $\text{WF}(\mathcal{P}_1 \cup \mathcal{P}_2 \cup \mathcal{P}_3 \cup \mathcal{P}_4 \cup \mathcal{P}_5) =$   
 $\text{WF}(\mathcal{P}_1 \cup \mathcal{P}_3) \cup \text{WF}(\mathcal{P}_4 \cup \mathcal{P}_2 \cup \mathcal{P}_3) \cup \text{WF}(\mathcal{P}_5 \cup \mathcal{P}_2 \cup \mathcal{P}_3).$   
Original search space size:  $2^{14}$

## Definition

Let an operator  $O$  over the *power set* lattice  $\bigotimes_{i \in I} \mathcal{L}_i$  and CIT  $T = (V, E, \nu)$  be given s.t.  $V_l$  are the leafs of  $T$ . The CIT-partition-size of  $O$  relative to  $(V, E, \nu)$  is defined as

$$\max(\{|\bigotimes_{i \in I_j} \mathcal{L}_i \otimes \bigotimes_{i \in I_3} \mathcal{L}_i| \mid v \in V_l, \nu(v) = \langle l_1, l_2, l_3 \rangle, j = 1, 2\})$$

## Definition

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## Proposition

Let a  $\leq_{\otimes}$ -monotonic operator  $O$  over the product lattice  $\bigotimes_{i \in I} \mathcal{L}_i$  and CIT  $T = (V, E, \nu)$  with CIT-partition-size  $s$  be given. Assume that  $O(x, y)$  can be computed to a call to an NP-oracle. The least fixpoint of  $O$  can be computed in time  $O(f(s))$ .



# Rounding Up

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## Other things in the paper

- Syntactical (sufficient) criteria for conditional independence in logic programs.
- Comparison with treewidth-based decompositions of logic programs.
- (Non)-satisfaction of graphoid properties.
- Comparison with Darwiche's logical notion of independence [Dar97].
- Relations with splitting in AFT [VGD06].

# Conclusion

- Algebraic account of conditional independence of sublattice w.r.t. an operator.
- Language-independent account of how the structure of problems in fixpoint-based logic allows to reduce global to parallel instances of local problems.
- Immediately applicable to logic program, formal argumentation, ADFs, default logic, ...

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**Thank you for your attention. Questions?**



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