



# Operator-based semantics for logic programs: recent advances

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Jesse Heyninck

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Open Universiteit, the Netherlands  
University of Cape Town, South Africa

# Why Operator-Based Semantics?

- Powerful framework for **defining and studying semantics rule-based languages**

If you have an idea of how to evaluate rule bodies in a three-valued setting, the framework does the rest of you.

- Single **unifying framework** that allows to derive most common logic programming semantics (and the semantics of default logic, autoepistemic logic, abstract argumentation, ...)
- Gives a **natural justification** of the well-founded and stable model semantics.
- Applied in **grounders** such as gringo.
- Useful for **explanations**.

# Goal of this talk

- Give an overview of the operator-based view on semantics of logic programming [VGRS91, Fit06, DMT00].
- Given an overview of some recent developments (by Bart, Ofer and myself) in this area:
  - Non-deterministic operators,
  - Disjunctive logic programs,
  - Choice logic programs.

# Goals and Structure

Syntax of Logic Programs

Semantics of Positive Programs

Semantics of Normal Logic Programs

Non-Deterministic Operators

Stable Semantics

- Stable Semantics for Deterministic Operators

- Stable Non-Deterministic Operators

Approximation Fixpoint Theory

Round up

# Syntax of Logic Programs

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# Syntax of Logic Programs

Set of atoms  $\mathcal{A} = \{a, b, c, p, q, r, a_1, a_2, \dots\}$

$$a \leftarrow b_1, \dots, b_n, \neg c_1, \dots, \neg c_m$$

- Program is a set of rules.
- Rule is positive if  $m = 0$ .
- Program is positive if all the rules are positive.

# Semantics of Positive Programs

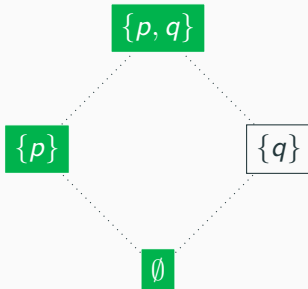
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# What are the semantics of logic programs?

$$p \leftarrow q.$$

Classical models?  $\emptyset, \{p\}, \{p, q\}$ .

Notice: a formula follows from every classical model if it follows from the minimal model  $\emptyset$ .





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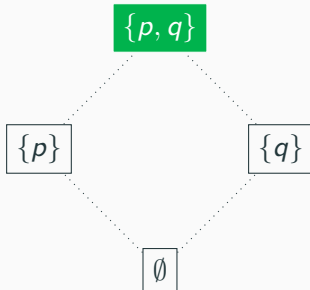
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# $T_{\mathcal{P}}$ -operator

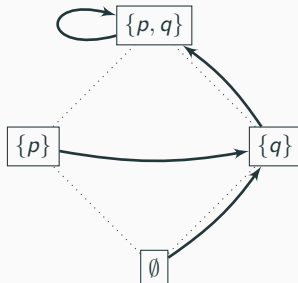
$$T_{\mathcal{P}} : \wp(\mathcal{AP}) \mapsto \wp(\mathcal{AP})$$

$$T_{\mathcal{P}}(\mathbf{x}) = \{a \mid a \leftarrow b_1, \dots, b_n \in \mathcal{P} \text{ and } b_1, \dots, b_n \in \mathbf{x}\}$$

## Example

$$\mathcal{P} = \{p \leftarrow q., \quad q \leftarrow .\}$$

$x$	$\emptyset$	$\{p\}$	$\{q\}$	$\{p, q\}$
$T_{\mathcal{P}}(x)$	$\{q\}$	$\{q\}$	$\{p, q\}$	$\{p, q\}$



# Models and Fixpoints

## Definition

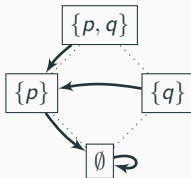
$x$  is a *pre-fixpoint* of  $T_{\mathcal{P}}$  if  $T_{\mathcal{P}}(x) \subseteq x$ .

*Intuition:* everything I can derive from  $x$  using  $\mathcal{P}$  is in  $x$ .

*Models* of  $\mathcal{P}$ .

## Example

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



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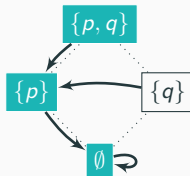
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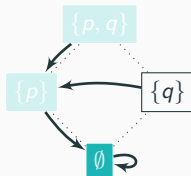
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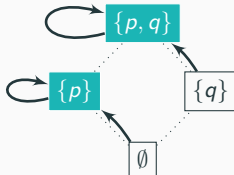
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## Example

$$\mathcal{P} = \{p \leftarrow \cdot, \quad q \leftarrow q.\}$$





# Models and Fixpoints

- For positive programs,  $T_{\mathcal{P}}$  has a unique **least fixpoint**  $x$ .
- It is also the least pre-fixpoint.
- We can compute it by iterating  $T_{\mathcal{P}}$  starting from  $\emptyset$ :  
$$T_{\mathcal{P}}(\dots T_{\mathcal{P}}(\emptyset) \dots) = \bigcup_{i \geq 0} T_{\mathcal{P}}^i(\emptyset)$$
  
And this is possible in polynomial time.

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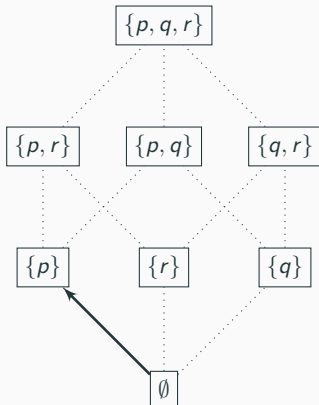
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Underlying result: A  $\subseteq$ -monotonic operator over a complete lattice admits a least fixpoint.

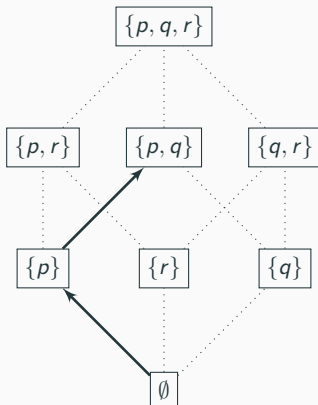
# Least fixpoint computation

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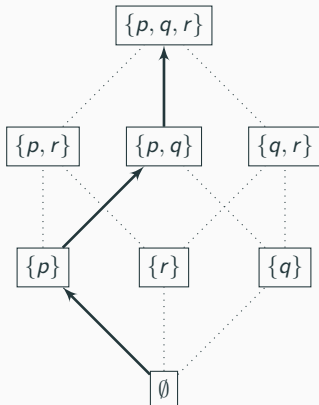
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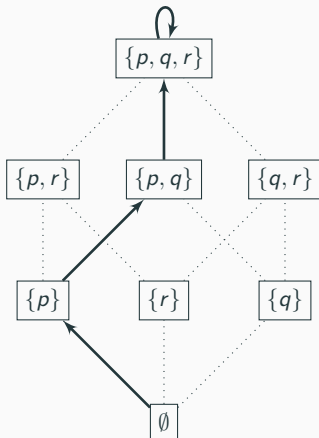
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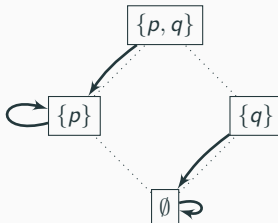
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# Least fixpoint $\neq$ unique fixpoint

Example ( $\mathcal{P} = \{p \leftarrow p.\}$ )





# Semantics of Normal Logic Programs

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$$p \leftarrow \neg q$$

How to extend our operator?

# Enters Negation

$$p \leftarrow \neg q$$

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Easy:

$$T_{\mathcal{P}}(x) = \{a \mid a \leftarrow b_1, \dots, b_n, \neg c_1, \dots, \neg c_m \in \mathcal{P}, \text{ and} \\ b_1, \dots, b_n \in x, \text{ and } c_1, \dots, c_m \notin x\}$$

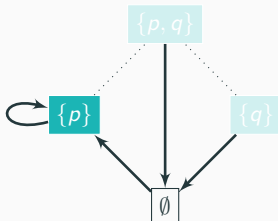
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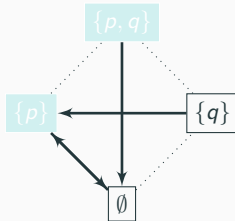
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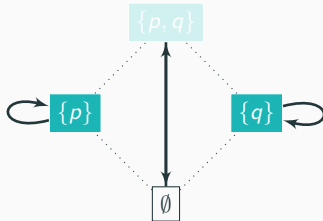
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$$\mathcal{P} = \{p \leftarrow \neg p\}$$



## No unique fixpoint

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



## Problems with negation

- There might not be a fixpoint.
- There might be multiple minimal fixpoints.
- We don't know how to find fixpoints.



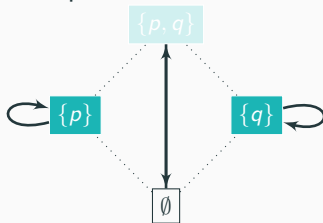
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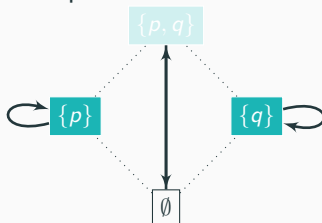
⇒ It is not a monotonic operator



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**Solution:** approximations of operators that are  $\leq_i$ -monotonic.

# Approximations

- Pairs of sets of atoms ( $x, y$ ).
  - $x$  contains all atoms that are definitely true.
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  - $(x_1, y_1) \leq_t (x_2, y_2)$  if  $x_1 \subseteq x_2$  and  $y_1 \subseteq y_2$ .
  - $(x_1, y_1) \leq_i (x_2, y_2)$  if  $x_1 \subseteq x_2$  and  $y_2 \subseteq y_1$ .

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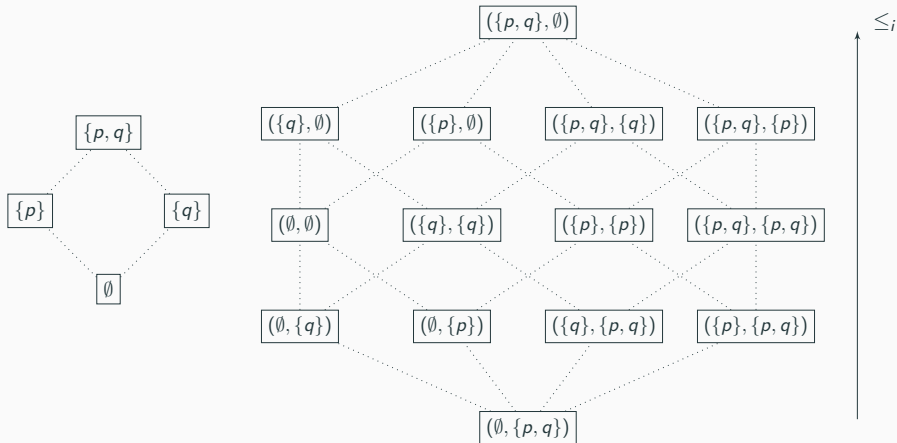
**Example (Given  $\mathcal{A} = \{p, q, r\}$ )**

$(\{p\}, \{p, q\})$ :  $p$  is true and  $q$  can be true.

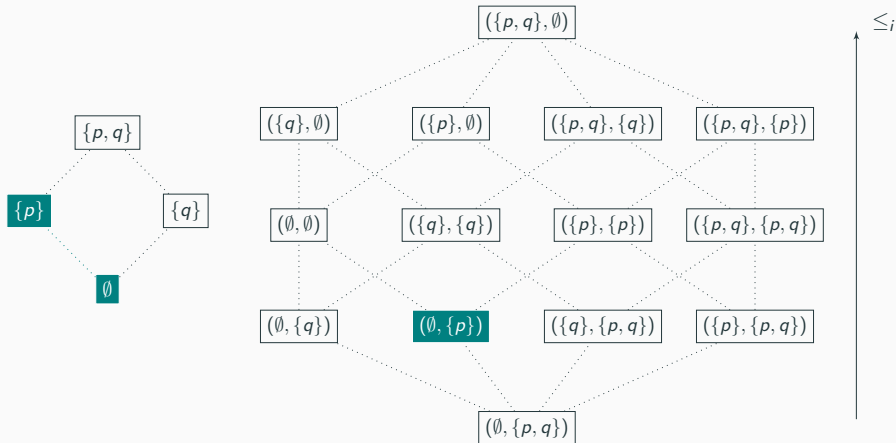
$$(\{p\}, \{p\}) \leq_t (\{p\}, \{p, q\})$$

$$(\{p\}, \{p, q\}) \leq_i (\{p\}, \{p\})$$

# Graphical Depiction

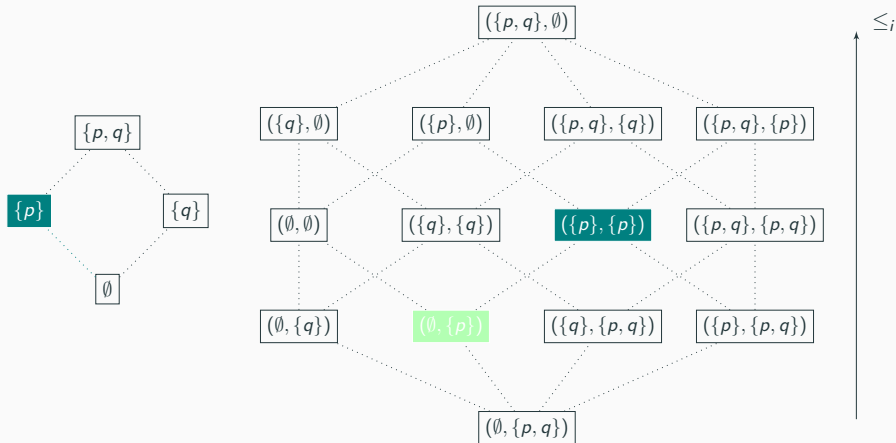


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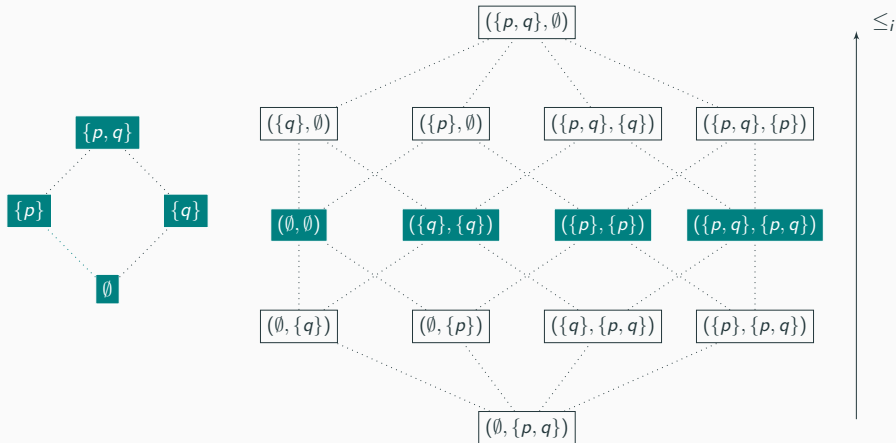




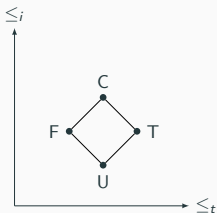
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# Approximations as Four-Valued Interpretations



- $-F = T, -T = F, -U = U$  and  $-C = C$
- $(x, y)(p) = \begin{cases} T & \text{if } p \in x \text{ and } p \in y, \\ U & \text{if } p \notin x \text{ and } p \in y, \\ F & \text{if } p \notin x \text{ and } p \notin y, \\ C & \text{if } p \in x \text{ and } p \notin y. \end{cases}$
- $(x, y)(\neg \phi) = -(x, y)(\phi),$
- $(x, y)(\psi \wedge \phi) = \text{lub}_{\leq_t} \{(x, y)(\phi), (x, y)(\psi)\},$
- $(x, y)(\psi \vee \phi) = \text{glb}_{\leq_t} \{(x, y)(\phi), (x, y)(\psi)\}.$

## Example

$(\{p\}, \{p, q\})(p) = T \quad (\{p\}, \{p, q\})(q) = U \quad (\{p\}, \{p, q\})(r) = F.$

$(\{p\}, \{p, q\})(\neg p) = F \quad (\{p\}, \{p, q\})(\neg q) = U$

$(\{p\}, \{p, q\})(p \wedge q) = U \quad (\{p\}, \{p, q\})(q \vee r) = U$

## Approximating $T_{\mathcal{P}}$ (from below)

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We input an approximation and output an approximation.

$$\mathcal{IC}_{\mathcal{P}}^l(x, y) = \{a \in \mathcal{A} \mid a \leftarrow b_1, \dots, b_n, \neg c_1, \dots, \neg c_m \in \mathcal{P}, \\ b_1, \dots, b_n \in x \text{ and } c_1, \dots, c_m \notin y\}$$

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**Example**  $(\{p \leftarrow p, \neg q\})$

$$\mathcal{IC}_{\mathcal{P}}^l(\{p\}, \{p, q\}) = \emptyset$$

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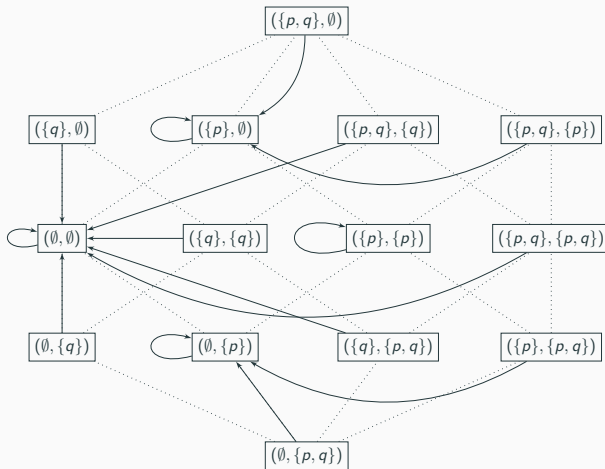
**Example**  $(\{p \leftarrow p, \neg q\})$

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# The Approximation Operator $\mathcal{IC}_{\mathcal{P}}$ (for $\mathcal{P} = \{p \leftarrow p, \neg q\}$ )

$$\mathcal{IC}_{\mathcal{P}}(x, y) = (\mathcal{IC}_{\mathcal{P}}^l(x, y), \mathcal{IC}_{\mathcal{P}}^u(x, y))$$



## Properties of $\mathcal{IC}_{\mathcal{P}}$

- $\mathcal{IC}_{\mathcal{P}}$  approximates  $T_{\mathcal{P}}$ :

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- $\mathcal{IC}_{\mathcal{P}}$  is  $\leq_i$ -monotonic:

if  $(x_1, y_1) \leq_i (x_2, y_2)$  then  $\mathcal{IC}_{\mathcal{P}}(x_1, y_1) \leq_i \mathcal{IC}_{\mathcal{P}}(x_2, y_2)$ .

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We say  $\mathcal{IC}_{\mathcal{P}}$  is an approximation operator. It is also symmetric, in the sense that  $\mathcal{IC}_{\mathcal{P}}(x, y) = (\mathcal{IC}_{\mathcal{P}}^l(x, y), \mathcal{IC}_{\mathcal{P}}^l(y, x))$ .

The  $\leq_i$ -monotonicity is our *indulgentia* back into Tarski's heaven:

## Proposition

$\mathcal{IC}_{\mathcal{P}}$  has a least fixpoint, obtainable as  $\bigsqcup_{i \geq 0} \mathcal{IC}_{\mathcal{P}}^i(\emptyset, \mathcal{A})$ <sup>1</sup>.

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# Kripke-Kleene Fixpoint

The  $\leq_i$ -monotonicity is our *indulgentia* back into Tarski's heaven:

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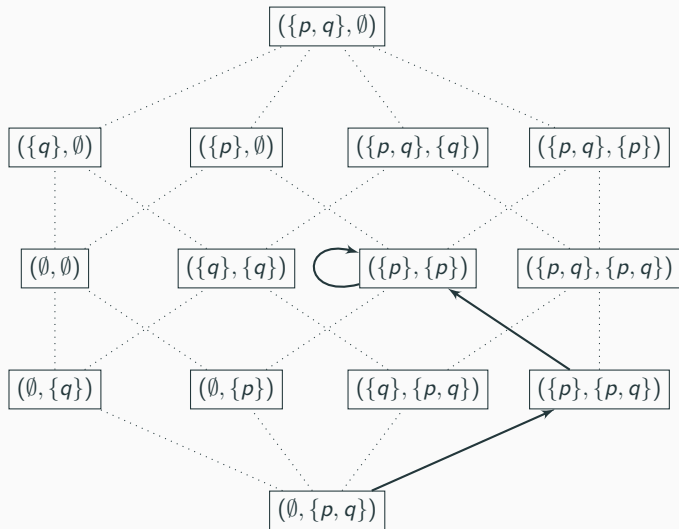
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If  $(x, y) = \mathcal{IC}_{\mathcal{P}}(x, y)$  then we call it a **partial supported model**.

---

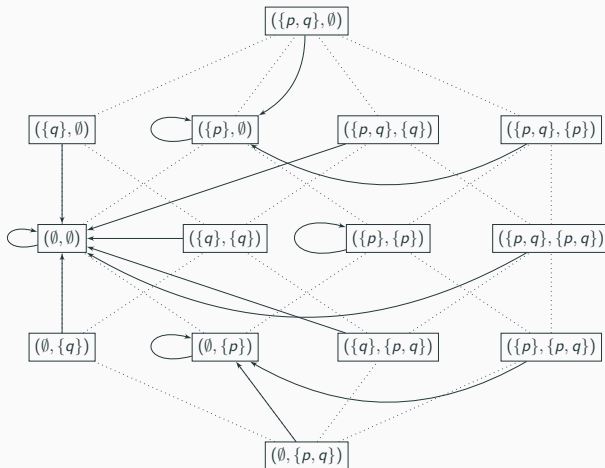
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**Example:**  $\mathcal{P} = \{p \leftarrow; q \leftarrow \neg p\}$



# The Approximation Operator $\mathcal{IC}_{\mathcal{P}}$ (for $\mathcal{P} = \{p \leftarrow p, \neg q\}$ )

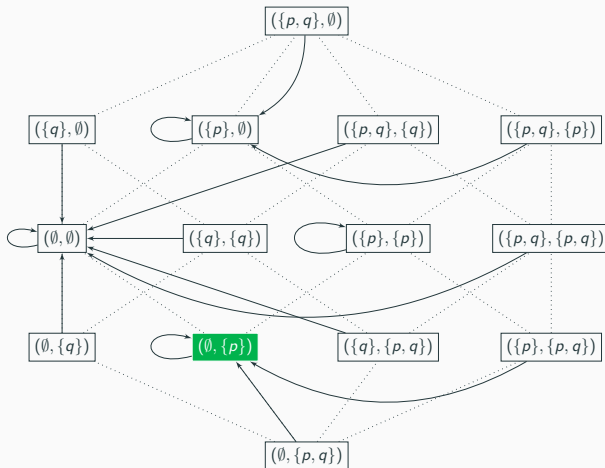
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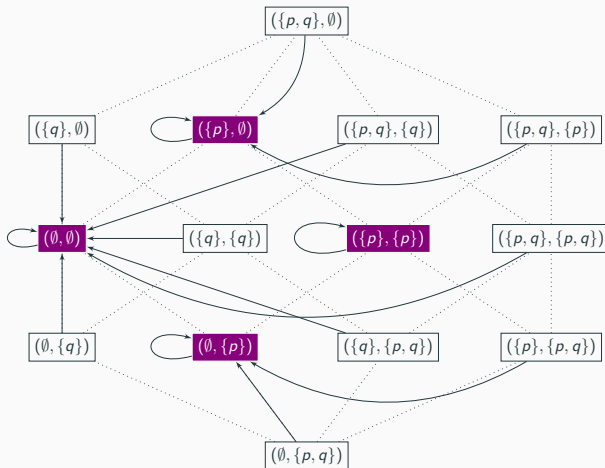
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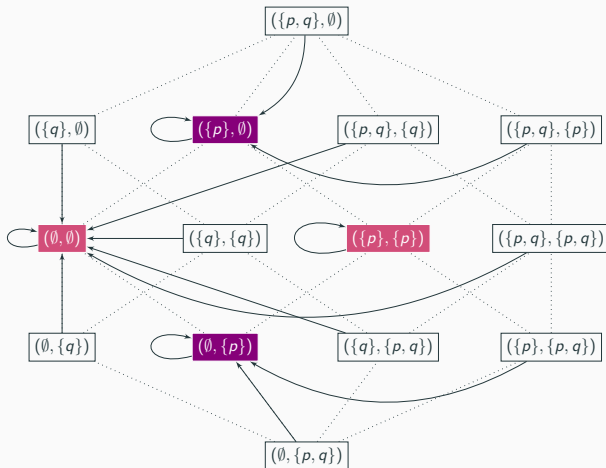




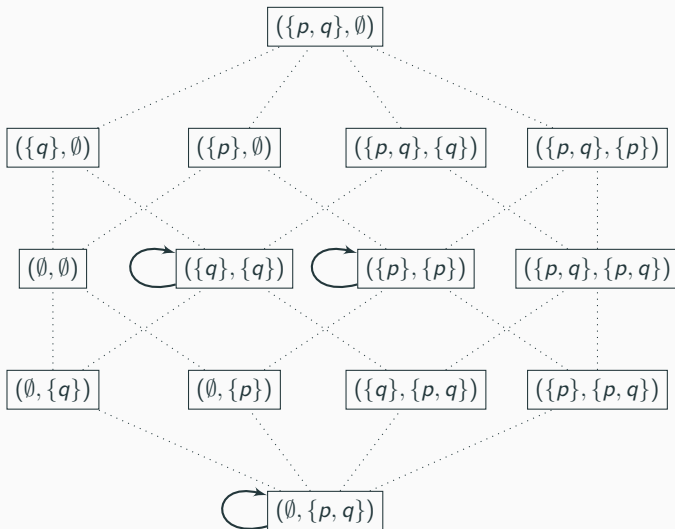
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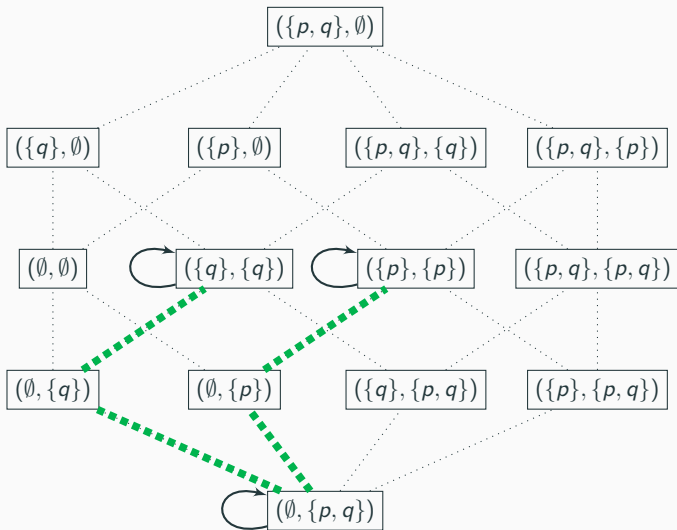
Supported models



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# Non-Deterministic Operators

---

# Motivation

```
{processed(M, T, 1..A): availability(M,T,A)} C :-  
    capacity(T,C).  
processed(M, T, N-1) :- processed(M, T, N), N>1.  
result(P,0,I) :- inventory(P,0,I), I>=0.  
result(P,0,-I) :- backlog(P,0,I), I>0.  
result(P,T,R1) :- demand(P,T,D), result(P,T-1,R), T>0,  
S=#sum{Y,M,X : processed(M,T,X), yield(M,P,Y)}, R1=S-D+R.
```

# Choice Atoms

A *choice atom* is an expression  $C = (\text{dom}, \text{sat})$  where  $\text{dom} \subseteq \mathcal{A}$  and  $\text{sat} \subseteq \wp(\text{dom})$ .

## Example

$1\{p, q, r\}2$  corresponds to the choice atom

$$C_1 = (\{p, q, r\}, \{\{p\}, \{q\}, \{r\}, \{p, q\}, \{p, r\}, \{q, r\}\}).$$

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- $\{p, q, s\}$  satisfies  $C_1$  as  $\{p, q, s\} \cap \{p, q, r\} = \{p, q\} \in \{\{p\}, \{q\}, \{r\}, \{p, q\}, \{p, r\}, \{q, r\}\}$ .
- $\{p, q, r\}$  does not satisfy  $C_1$  as  $\{p, q, r\} \cap \{p, q, r\} = \{p, q, r\} \notin \{\{p\}, \{q\}, \{r\}, \{p, q\}, \{p, r\}, \{q, r\}\}$ .



## Choice Rules

Where  $C, C_1, \dots, C_n$  are choice atoms, a **choice rule** is of the form:

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- A choice rule is **normal** if  $\text{sat}(C_i) = \{\{a\}\}$  (for some  $a \in \mathcal{A}$ ) or  $\text{sat}(C_i) = \{\emptyset\}$  for every  $i = 1 \dots n$ ,
  - $(\{a\}, \{\{a\}\})$  is denoted by  $a$ , and
  - $(\{a\}, \{\emptyset\})$  is denoted by  $\neg a$ .

# Non-Deterministic Immediate Consequence Operator

Given a choice program  $\mathcal{P}$  and a set of atoms  $x$ :

- A rule  $r \in \mathcal{P}$  is  **$x$ -applicable** (in symbols,  $r \in \mathcal{P}(x)$ ) if  $x$  satisfies the body of  $r$ .
- **$IC_{\mathcal{P}}(x)$**  =  $\{z \subseteq \bigcup_{r \in \mathcal{P}(x)} \text{dom}(\text{hd}(r)) \mid \forall r \in \mathcal{P}(x) : z \models \text{hd}(r)\}$

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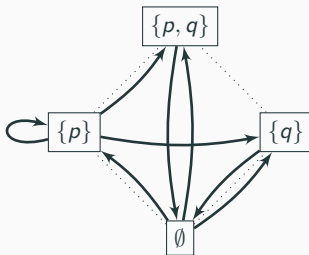
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# Non-Deterministic Approximators for Normal Choice Programs

$$\mathcal{HD}_{\mathcal{P}}^l(x, y) = \{C \mid C \leftarrow a_1, \dots, a_n, \neg b_1, \dots, \neg b_m \in \mathcal{P}, \\ \{a_1, \dots, a_n\} \subseteq x, \{b_1, \dots, b_m\} \cap y = \emptyset\}$$

$$\mathcal{IC}_{\mathcal{P}}^l(x, y) = \{z \subseteq \bigcup_{C \in \mathcal{HD}_{\mathcal{P}}^l(x, y)} \text{dom}(C) \mid \forall C \in \mathcal{HD}_{\mathcal{P}}^l(x, y), z \cap \text{dom}(C) \in \text{sat}(C)\}$$

$$\mathcal{IC}_{\mathcal{P}}^u(x, y) = \mathcal{IC}_{\mathcal{P}}^l(y, x)$$

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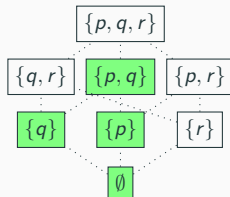
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$$\mathcal{IC}_{\mathcal{P}}^u(\mathbf{x}, \mathbf{y}) = \mathcal{IC}_{\mathcal{P}}^l(\mathbf{y}, \mathbf{x})$$

$$\mathcal{IC}_{\mathcal{P}}(\mathbf{x}, \mathbf{y}) = (\mathcal{IC}_{\mathcal{P}}^l(\mathbf{x}, \mathbf{y}), \mathcal{IC}_{\mathcal{P}}^u(\mathbf{x}, \mathbf{y}))$$

**Example:**  $\mathcal{P} = \{1\{p, q\}2 \leftarrow \neg r.\}$

- $\mathcal{HD}_{\mathcal{P}}^{c,l}(\emptyset, \{r\}) = \emptyset,$
- $\mathcal{HD}_{\mathcal{P}}^{c,u}(\emptyset, \{r\}) = \{1\{p, q\}2\},$
- $\mathcal{IC}_{\mathcal{P}}^c(\emptyset, \{r\}) = (\{\emptyset\}, \{\{p\}, \{q\}, \{p, q\}\}).$





# Approximators for Choice Programs (aka the real fun)

Given a choice program  $\mathcal{P}$  and pair of sets of atoms  $(x, y)$  let:

$$\begin{aligned}\mathcal{HD}_{\mathcal{P}}^{\text{GZ},l}(x, y) &= \{C \mid \exists C \leftarrow C_1, \dots, C_n \in \mathcal{P}, \forall i = 1 \dots n : \\ &\quad x \cap \text{dom}(C_i) = y \cap \text{dom}(C_i) \in \text{sat}(C_i)\}, \\ \mathcal{HD}_{\mathcal{P}}^{\text{LPST},l}(x, y) &= \{C \mid \exists C \leftarrow C_1, \dots, C_n \in \mathcal{P}, \forall i = 1 \dots n : \\ &\quad \forall z \in [x, y] : z(C_i) = \text{T}\}, \\ \mathcal{HD}_{\mathcal{P}}^{\text{MR},l}(x, y) &= \{C \mid \exists C \leftarrow C_1, \dots, C_n \in \mathcal{P}, \exists z \subseteq x : \\ &\quad \forall i = 1 \dots n : y(C_i) = \text{T} \text{ and } z(C_i) = \text{T}\},\end{aligned}$$

For  $x \in \{\text{LPST}, \text{MR}, \text{GZ}\}$  let:

$$\mathcal{IC}^{x,l}(x, y) = \{z \subseteq \bigcup_{C \in \mathcal{HD}_{\mathcal{P}}^{x,l}(x, y)} \text{dom}(C) \mid \forall C \in \mathcal{HD}_{\mathcal{P}}^{x,l}(x, y) : z \cap \text{dom}(C) \in \text{sat}(C)\}$$

For  $\dagger \in \{\text{MR}, \text{LPST}, \mathcal{U}\}$  let:

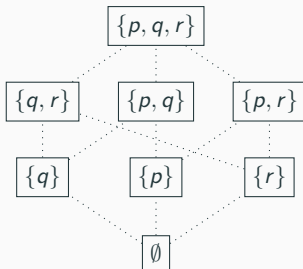
$$\mathcal{IC}_{\mathcal{P}}^{\mathcal{U},l}(x, y) = \mathcal{IC}_{\mathcal{P}}^{\dagger,u}(x, y) = \bigcup_{x \subseteq z \subseteq y} \mathcal{IC}_{\mathcal{P}}(z)$$

whereas  $\mathcal{IC}_{\mathcal{P}}^{\text{GZ},u}(x, y) = \mathcal{IC}_{\mathcal{P}}^{\text{GZ},l}(x, y)$ .  $\mathcal{IC}^x(x, y) = (\mathcal{IC}^{x,l}(x, y), \mathcal{IC}^{x,u}(x, y))$  (for  $x \in \{\text{LPST}, \text{MR}, \text{GZ}, \mathcal{U}\}$ ).

## Intermezzo on Orders

Let  $L = \langle \mathcal{L}, \leq \rangle$  be a lattice and  $X, Y \in 2^{\mathcal{L}}$ .

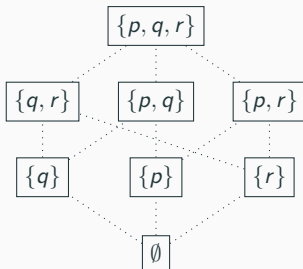
- $X_1 \preceq_L^S X_2$  if for every  $x_2 \in X_2$  there is an  $x_1 \in X_1$  s.t.  $x_1 \leq x_2$ .
- $Y_1 \preceq_L^H Y_2$  if for every  $y_1 \in Y_1$  there is a  $y_2 \in Y_2$  s.t.  $y_1 \leq y_2$ .
- $(X_1, Y_1) \preceq_i^A (X_2, Y_2)$  iff  $X_1 \preceq_L^S X_2$  and  $Y_2 \preceq_L^H Y_1$ .



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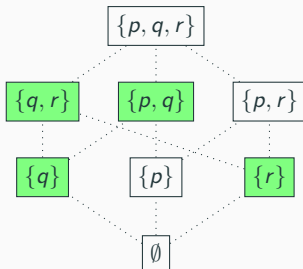


Pairs of sets  $(X, Y)$  as convex sets.

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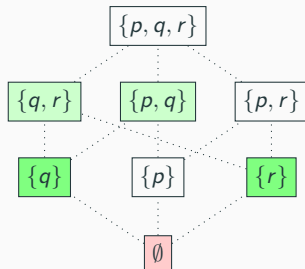


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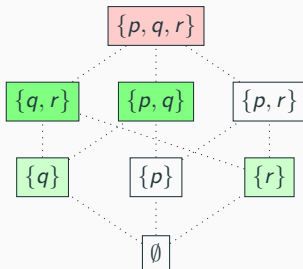


$$\{\emptyset\} \preceq_L^S \{\{q\}, \{r\}\}$$

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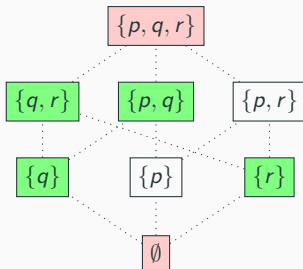


$$\{\{q, r\}, \{p, q\}\} \preceq_L^H \{\{p, q, r\}\}$$

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$$(\{\emptyset\}, \{\{p, q, r\}\}) \preceq_i^A (\{\{q\}, \{r\}\}, \{\{q, r\}, \{p, q\}\})$$

# Non-Deterministic Approximation Operators

$O$  is approximated using a non-deterministic approximation operator  $\mathcal{O} : \mathcal{L}^2 \rightarrow 2^{\mathcal{L}} \times 2^{\mathcal{L}}$ :

- that is  $\preceq_i^A$ -monotonic,
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Output of  $\mathcal{O}(x, y) = (\{x_1, x_2, \dots\}, \{y_1, y_2, \dots\})$  is a set of lower bounds respectively upper bounds on choices  $O(z) = \{z_1, z_2, \dots\}$  (with  $x \leq z \leq y$ ).

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Output of  $\mathcal{O}(x, y) = (\{x_1, x_2, \dots\}, \{y_1, y_2, \dots\})$  is a set of lower bounds respectively upper bounds on choices  $O(z) = \{z_1, z_2, \dots\}$  (with  $x \leq z \leq y$ ).

## Non-Deterministic AFT: fixpoints

$\mathcal{IC}_{\mathcal{P}}$  allows to generalize supported models to the three-valued case:

**Example**  $\mathcal{P} = \{\{p, q\} = 1 \leftarrow \neg p. \quad p \leftarrow q.\}$

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- No (two-valued) supported models
- $\mathcal{IC}_{\mathcal{P}}^c(\emptyset, \{p\}) = (\{\emptyset\}, \{\{p\}, \{q\}, \{p, q\}\})$ .
- (among others)

## Proposition

*Let a normal logic program  $\mathcal{P}$  be given. Then  $(x, y) \in \mathcal{IC}_{\mathcal{P}}(x, y)$  iff  $(x, y)$  is a partial supported model of  $\mathcal{P}$ .*

## Proposition

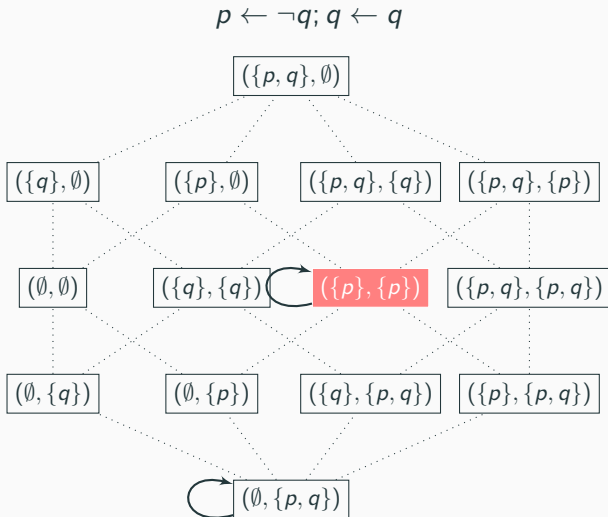
*Let a choice program  $\mathcal{P}$  and  $x \in \{\text{LPST}, \text{MR}, \text{GZ}, \mathcal{U}\}$  be given. If  $(x, y) \in \mathcal{IC}_{\mathcal{P}}^x(x, y)$  then for every  $a \in y$ , there is a  $C \leftarrow C_1, \dots, C_n$  with  $a \in \text{dom}(C)$  s.t.*

$$\mathcal{IC}_{\{p \leftarrow C_1, \dots, C_n\}}^x(x, y) = (\{\{p\} \cap x\}, \{\{p\} \cap y\}).$$

# Stable Semantics

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Construction of the Kripke-Kleene fixpoint:

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- Fixpoint reached.

Can't get rid of the self-supporting atom  $q$  in the upper bound.

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$S(\mathcal{IC}_{\mathcal{P}}^l)(\{p, q\}) = \emptyset$  since:

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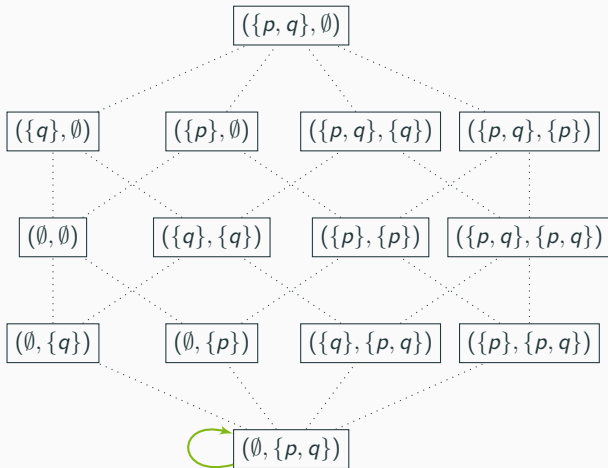
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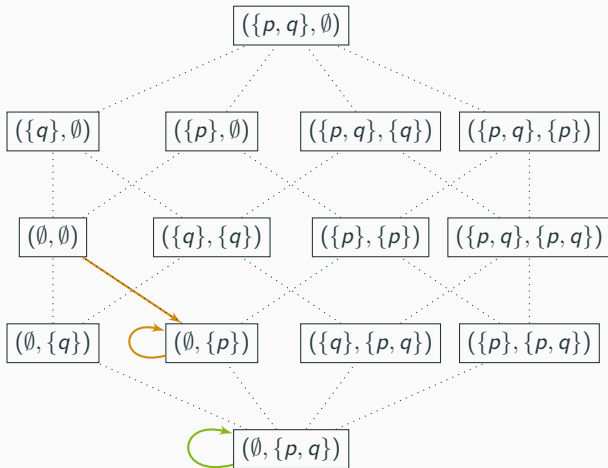
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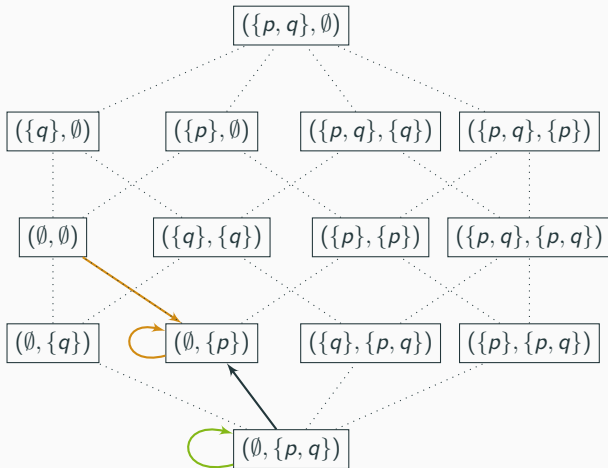
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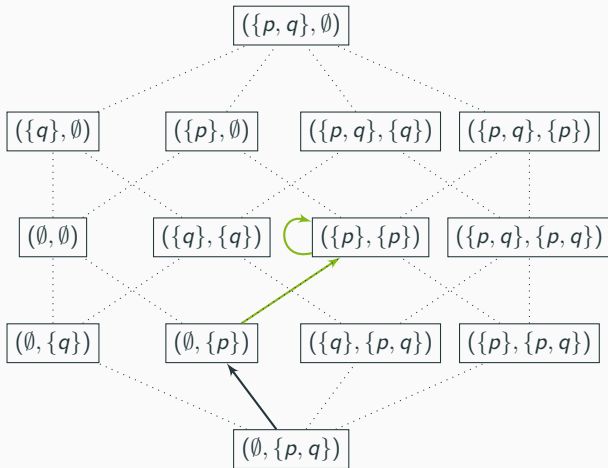
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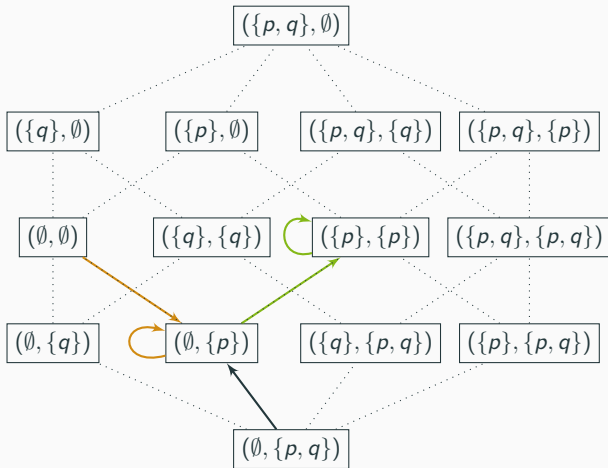
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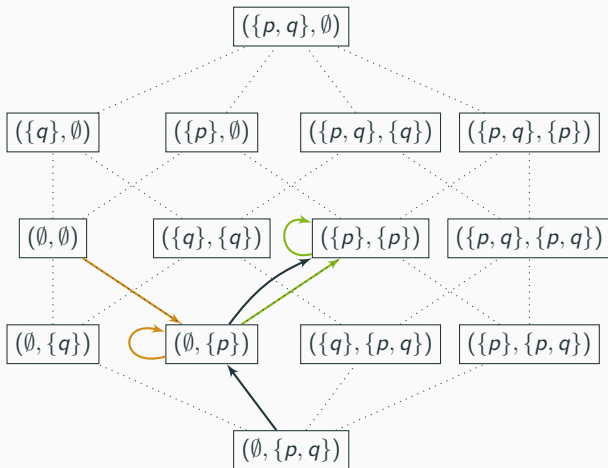
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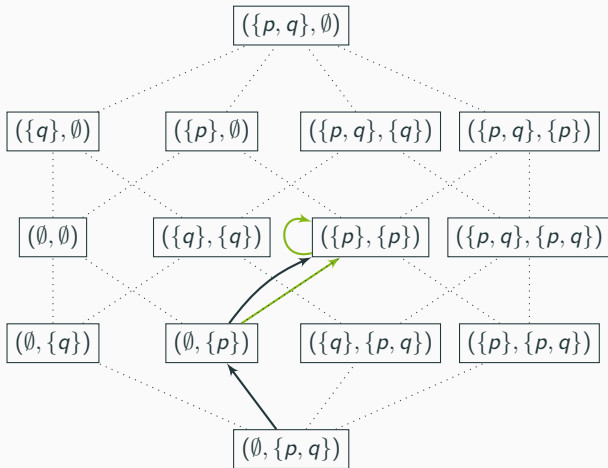
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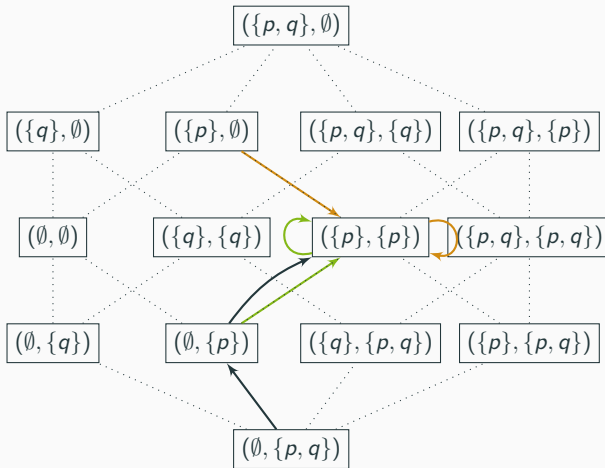
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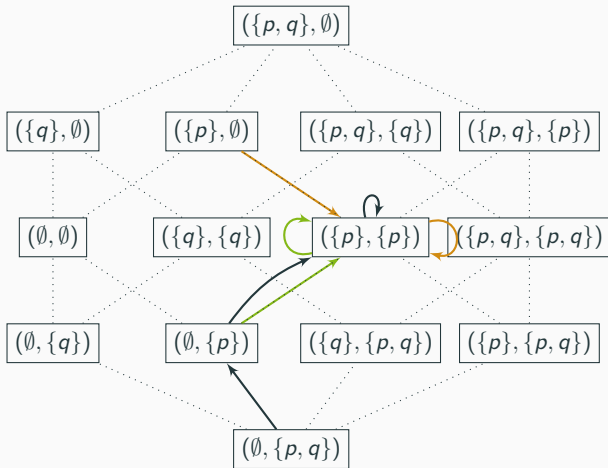
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# Stable Operator and Well-Founded Model

$$S(IC_{\mathcal{P}}^l)(y) = lfp(IC_{\mathcal{P}}^l(\cdot, y)) \quad S(IC_{\mathcal{P}}^u)(x) = lfp(IC_{\mathcal{P}}^u(x, \cdot))$$

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If  $(x, y) = S(IC_{\mathcal{P}})(x, y)$ , we call it a **(partial) stable model**.

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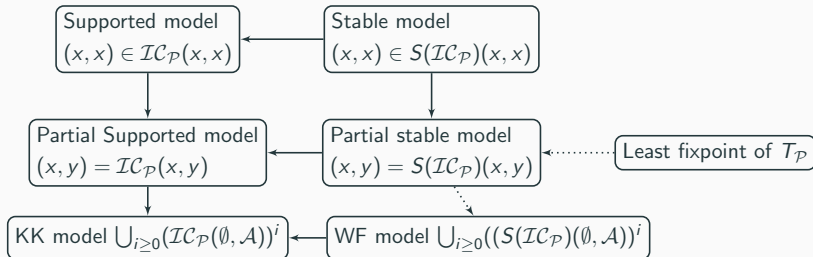
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- If  $T_{\mathcal{P}}$  has a least fixpoint, it coincides with the well-founded model.

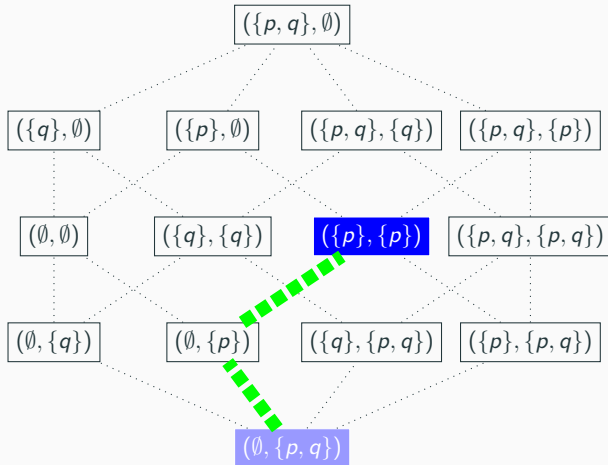
# Taking Stock

Semantics	= Stable?	+ Operator	+ Fixpoint
	$S$	$(\mathcal{O})(x, y)$	$= (x, y)$
Kripke-Kleene	No	$\mathcal{IC}_{\mathcal{P}}$	least fp
Partial Supported	No	$\mathcal{IC}_{\mathcal{P}}$	any fp
Supported	No	$\mathcal{IC}_{\mathcal{P}}$	total fp ( $x = y$ )
Well-founded	yes	$\mathcal{IC}_{\mathcal{P}}$	least fp
Partial Stable	yes	$\mathcal{IC}_{\mathcal{P}}$	any fp
Stable	yes	$\mathcal{IC}_{\mathcal{P}}$	total fp ( $x = y$ )



# Stable Operator: Example

$$p \leftarrow \neg q; q \leftarrow q$$



## Stable Operator: Example 2

$$\mathcal{P} = \{p \leftarrow \neg q; \quad q \leftarrow \neg p; \quad r \leftarrow r; \quad s \leftarrow \neg r\}$$

- Kripke-Kleene fixpoint:  $(\emptyset, \{p, q, r, s\})$ .
- Well-founded model:  $(\{s\}, \{p, q, s\})$ .
- Stable models:  $(\{p, s\}, \{p, s\}), (\{q, s\}, \{q, s\})$ .

$$\frac{\mathcal{P}}{x} = \{a \leftarrow b_1, \dots, b_n \mid a \leftarrow b_1, \dots, b_n, \neg c_1, \dots, \neg c_m \in \mathcal{P} \\ c_1, \dots, c_n \notin x\}$$

## Definition

$x$  is a *stable model* of  $\mathcal{P}$  if it is a minimal model of  $\frac{\mathcal{P}}{x}$ .

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**Example** ( $\mathcal{P} = \{p \leftarrow \neg p; q \leftarrow \neg p; p \leftarrow \neg q\}$ )

$\frac{\mathcal{P}}{\{q\}} = \{p \leftarrow; q \leftarrow\}$ .  $\{q\}$  is not a minimal model of  $\mathcal{P}$ , thus  $\{q\}$  is not a stable model.

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## Proposition

$S(\mathcal{IC}_{\mathcal{P}}^l)(y)$  is the set of minimal models of  $\frac{\mathcal{P}}{y}$ .

## Proposition

$(x, x) = S(\mathcal{IC}_{\mathcal{P}})(x, x)$  if and only if  $x$  is a stable model of  $\mathcal{P}$  (iff  $x = S(\mathcal{IC}_{\mathcal{P}}^l)(x)$ ).

## Stable Operator for Deterministic Approximation Operators

Basic idea: look for smallest fixpoint that the upper bound allows us to construct:

$$S(\mathcal{O}_I)(y) = glb\{x \in \mathcal{L} \mid x = \mathcal{O}_I(x, y)\}$$

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## Stable Operator: Example

$$\mathcal{P} = \{1\{p, q\}2 \leftarrow .\}$$

$x$	$\emptyset$	$\{p\}$	$\{q\}$	$\{p, q\}$
$\mathcal{IC}_{\mathcal{P}}^l(x, y)$	$\{p\}, \{q\}, \{p, q\}$	$\{p\}, \{q\}, \{p, q\}$	$\{p\}, \{p\}, \{p, q\}$	$\{p\}, \{q\}, \{p, q\}$

1. glb of fixpoints of  $\mathcal{IC}_{\mathcal{P}}^{c,l}(\cdot, y)$ :  $\{p\} \cap \{q\} \cap \{p, q\} = \emptyset$ .
2. Minimal fixpoints:  $\{p\}$  and  $\{q\}$ .
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We have two choices now: minimality based or iterative.



# Minimality-Based Stable Operator

Let an ndao  $\mathcal{O} : \mathcal{L}^2 \rightarrow 2^{\mathcal{L}} \times 2^{\mathcal{L}}$  and some  $x, y \in \mathcal{L}$  be given. Then:

- the **m-complete lower stable operator** as:

$$C^m(\mathcal{O}_l)(y) = \{x \in \mathcal{L} \mid x \in \mathcal{O}_l(x, y) \text{ and } \neg \exists x' < y : x' \in \mathcal{O}_l(x', y)\}$$

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- a **m-stable fixpoint** of  $\mathcal{O}$  as any  $(x, y) \in \mathcal{L}^2$  s.t.  
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$(\{p\}, \{p\})$  is a *m-stable fixpoint*.

**Theorem** Let a dlp  $\mathcal{P}$  and a consistent  $x \subseteq y \subseteq \mathcal{A}_{\mathcal{P}}^2$  be given.

Then  $(x, y)$  is a stable model of  $\mathcal{P}$  according to [Prz91] (i.e. the minimal models of the reduct) iff  $(x, y) \in S(\mathcal{IC}_{\mathcal{P}})(x, y)$ .

## Constructive Stable Operator

Given a non-deterministic operator  $O : \mathcal{L} \rightarrow \wp(\mathcal{L})$ , a sequence  $x_0, \dots, x_n \subseteq \mathcal{L}$  is **well-founded relative to  $O$**  if:

- $x_0 = \perp$ ,
- $x_i \leq x_{i+1}$  and  $x_{i+1} \in O(x_i)$  for every successor ordinal  $i \geq 0$ .
- $x_\lambda = (\text{lub}\{x_i\}_{i < \lambda})$  for a limit ordinal  $\lambda$ .

# Constructive Stable Operator

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The **complete constructive lower bound operator** is defined as:

$$S^c(IC_P^l)(y) = \{x \in \mathcal{O}_l(x, y) \mid \exists x_0, \dots, x \in \text{wfs}(IC_P^l(., y))\}$$

The complete constructive upper bound operator is defined analogously, and the constructive stable operator is defined as

$$S^c(IC_P)(x, y) = (S^c(IC_P^l)(y), S^c(IC_P^u)(x)).$$

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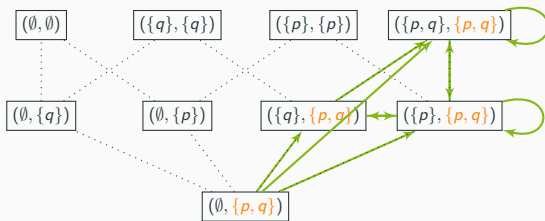
A pair  $(x, y)$  is a **constructive stable fixpoint** iff

$$(x, y) \in S^c(\mathcal{IC}_P)(x, y).$$

# Stable Operator: Example 1

$$\mathcal{P} = \{1\{p, q\}2 \leftarrow ., \quad p \leftarrow q\}$$

$y$	$\emptyset$	$\{p\}$	$\{q\}$	$\{p, q\}$
$IC_{\mathcal{P}}^{c,l}(\cdot, y)$	$\{p\}, \{q\}, \{p, q\}$	$\{p\}, \{q\}, \{p, q\}$	$\{p\}, \{p, q\}$	$\{p\}, \{p, q\}$



$$S(IC_{\mathcal{P}}^l)(\{p, q\}) =$$

$$\{\{p\}, \{p, q\}\}$$

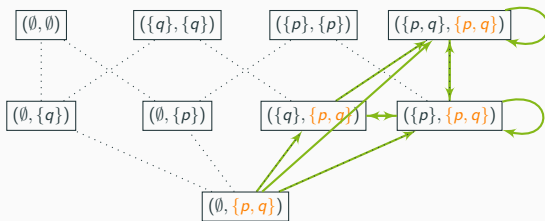
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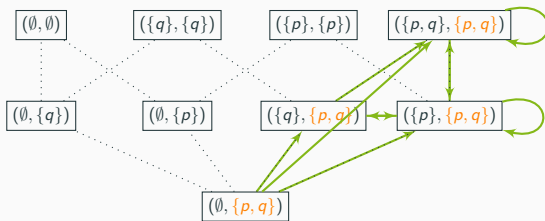
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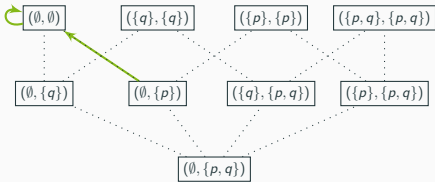
$$\{\{p\}, \{p, q\}\}$$

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$(\{p, q\}, \{p, q\})$  is a stable fp.

## Stable Operator: Example 2

$$\mathcal{P} = \{1\{p, q\} \leftarrow p.; \quad 1\{p, q\} \leftarrow q\}$$



Well-founded sequence of  $IC_{\mathcal{P}}^l(., y)$  for any  $y \subseteq \{p, q\}$ :  $\emptyset$

$(\emptyset, \emptyset)$  is the unique fixpoint of  $S(\mathcal{IC}_{\mathcal{P}})$ .

# General results and characterisation theorem

## Theorem

Let a choice program  $\mathcal{P}$  s.t. for every  $C_1 \leftarrow C_2, \dots, C_n \in \mathcal{P}$ ,  $\text{dom}(C_i)$  is finite for  $i = 1 \dots n$ , and  $x \in \{\text{MR}, \text{LPST}, \mathcal{U}\}$  and  $x \subseteq y \subseteq \mathcal{A}$  be given. Then  $S^c(\mathcal{IC}_{\mathcal{P}}^x)(x, y) \neq \emptyset$ . Furthermore,  $C^c(\mathcal{IC}_{\mathcal{P}}^{\text{GZ}, l})(y) \neq \emptyset$ .

## Theorem

Let a choice program  $\mathcal{P}$  be given.

1.  $x$  is a stable model according to [LPST10] iff  $(x, x)$  is a stable fixpoint of  $\mathcal{IC}_{\mathcal{P}}^{\text{LPST}}$ .
2.  $x$  is a stable model according to [MR04] iff  $(x, x)$  is a stable fixpoint of  $\mathcal{IC}_{\mathcal{P}}^{\text{MR}}$ .
3. If  $\mathcal{P}$  is a aggregate program then  $x$  is a stable model according to [GZ14] iff  $x \in C^c(\mathcal{IC}_{\mathcal{P}}^{\text{GZ}, l})(x)$ .

## Disjunctions are Choice Constructs

$$\text{D2C}(\mathcal{P}) = \{1\Delta \leftarrow \phi \mid \bigvee \Delta \leftarrow \phi \in \mathcal{P}\}.$$

**Example**  $\text{D2C}(\{p \vee q \leftarrow .\}) = \{1\{p \vee q\} \leftarrow .\}.$

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## Proposition

*For any disjunctive logic program  $\mathcal{P}$ ,  $\mathcal{IC}_{\mathcal{P}} = \mathcal{IC}_{\text{D2C}(\mathcal{P})}^c$ .*

# Disjunctions are Choice Constructs

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## Proposition

For any disjunctive logic program  $\mathcal{P}$ ,  $\mathcal{IC}_{\mathcal{P}} = \mathcal{IC}_{\text{D2C}(\mathcal{P})}^c.$

Difference between DLP and choice programs: the difference is *not* in the treatment of disjunction and choice atoms (i.e. when they should be made true or false), but rather in how the stable semantics is constructed:

- Disjunctions: minimality-based stable semantics [?]

$$S(\mathcal{O}_u)(x) = \{y \in \mathcal{L} \mid y \in \mathcal{O}_u(x, y) \text{ and } \neg \exists y' < y : y' \in \mathcal{O}_u(x, y')\}$$

$\{p\}$  and  $\{q\}$  are stable

- Choice constructs: constructive stable semantics

$\{p\}$ ,  $\{p, q\}$  and  $\{q\}$  are stable

# Taking Stock: Non-Deterministic Operators

Semantics	=	Stable? $S$	+	Operator $(\mathcal{O})(x, y)$	+	Fixpoint $\exists (x, y)$
Partial Supported		No		$\mathcal{IC}_{\mathcal{P}}$		any fp
Supported		No		$\mathcal{IC}_{\mathcal{P}}$		total fp ( $x = y$ )
Partial Stable for DLPs		M-stable		$\mathcal{IC}_{\mathcal{P}}$		any fp
Stable for DLPs		M-stable		$\mathcal{IC}_{\mathcal{P}}$		total fp ( $x = y$ )
Partial Stable for CPs		C-stable		$\mathcal{IC}_{\mathcal{P}}$		any fp
Stable for CPs		C-stable		$\mathcal{IC}_{\mathcal{P}}$		total fp ( $x = y$ )

# Taking Stock: Non-Deterministic Operators

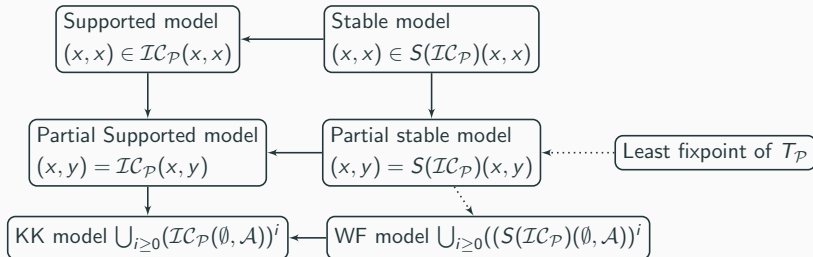
Semantics	=	Stable? $S$	+ Operator $(\mathcal{O})(x, y)$	+ Fixpoint $\ni (x, y)$
Partial Supported		No	$IC_{\mathcal{P}}$	any fp
Supported		No	$IC_{\mathcal{P}}$	total fp ( $x = y$ )
Partial Stable for DLPs		M-stable	$IC_{\mathcal{P}}$	any fp
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Stable for CPs		C-stable	$IC_{\mathcal{P}}$	total fp ( $x = y$ )
[LPST10]		C-stable	$IC_{\mathcal{P}}^{\text{LPST}}$	any fp
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		C-stable	$IC_{\mathcal{P}}^{\text{MR}}$	any fp
[MR04]		C-stable	$IC_{\mathcal{P}}^{\text{MR}}$	total fp ( $x = y$ )
		C-stable	$IC_{\mathcal{P}}^{\text{GZ}}$	any fp
[GZ14]		C-stable	$IC_{\mathcal{P}}^{\text{GZ}}$	total fp ( $x = y$ )



# Approximation Fixpoint Theory

---

# Recap



- Operator-based framework
  - Non-monotonic operator  $T_{\mathcal{P}}$ ,
  - a  $\leq_i$ -monotonic approximation operator  $\mathcal{IC}_{\mathcal{P}}$ ,
  - and its stable variant  $S(\mathcal{IC}_{\mathcal{P}})$ .
- Allows us to define semantics as fixpoints of these operators, with attractive properties.
- Generalized to the non-deterministic setting, and with other operators.
- This story can be told for a great number of formalisms

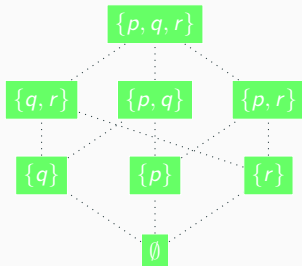
- State semantics for non-deterministic operators:
  - Kripke-Kleene state,
  - Well-founded state (relative to a stable construction),
  - Well-founded state with CWA (relative to a stable construction),
- Semi-equilibrium semantics,
- “Strongly” supported models,
- Regular models,
- M-Stable models,
- L-stable models.

# Well-founded State

Basic Idea: generalize operator  $\mathcal{O} : \mathcal{L}^2 \rightarrow 2^{\mathcal{L}} \times 2^{\mathcal{L}}$  to an operator  $\mathcal{O}' : 2^{\mathcal{L}} \times 2^{\mathcal{L}} \rightarrow 2^{\mathcal{L}} \times 2^{\mathcal{L}}$  as follows:

$$\mathcal{O}'((X, Y)) = \left( \bigcup_{x \in X, y \in Y} \mathcal{O}_l(x, y) \uparrow, \bigcup_{x \in X, y \in Y} \mathcal{O}_u(x, y) \downarrow \right)$$

**Example**  $\mathcal{P} = \{1\{p, q\} \leftarrow .; q \leftarrow \neg r.\}$ .

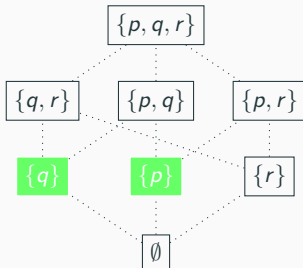


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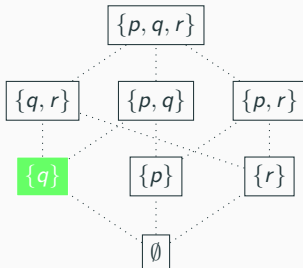


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**Example**  $\mathcal{P} = \{1\{p, q\} \leftarrow .; q \leftarrow \neg r.\}$ .



Given an ndao  $\mathcal{O}$  approximating a non-deterministic operator  $O$ , a pair  $(x, y)$  is a HT-*pair* (denoted  $(x, y) \in \text{HT}(\mathcal{O})$ ) if the following three conditions are satisfied:

- $x \leq y$ ,
- $O(y) \preceq_L^S y$ , and
- $\mathcal{O}_I(x, y) \preceq_L^S x$ .

**Proposition** Let some normal disjunctive logic program  $\mathcal{P}$  be given. Then:  $\text{HT}(\mathcal{P}) = \text{HT}(\mathcal{IC}_{\mathcal{P}})$ .

**Proposition** Given an upwards coherent ndao  $\mathcal{O}$ , (1) if  $(x, x) \in \mathcal{O}(x, x)$  then  $(x, x) \in \text{HT}(\mathcal{O})$ ; and (2)  $(x, x) \in \min_{\leq_t}(\text{HT}(\mathcal{O}))$  iff  $(x, x) \in S(\mathcal{O})(x, x)$ .

# Operator-Based Semantics for Dialects of Logic Programming

- ✓ Aggregates in the body:  $p \leftarrow \#sum\{2 : p; q : 1; r : 1\} \geq 2$ .
- ✓ Propositional formulas in the body:  $p \leftarrow q \wedge (r \vee (s \wedge \neg t))$ .
- ✓ Disjunctions in the head:  $p \vee q \leftarrow q \wedge (r \vee (s \wedge \neg t))$ .
- ✓ Choice constructs in the head:  $\#count\{p; q; r\} = 2 \leftarrow \neg r$ .
- ✓ DL-based logic programs:  $\mathbf{K}C(x) \leftarrow \neg p(X); C \sqsubseteq D$ .
- ✓ Higher-order logic programs:  $S(P, Q) \leftarrow; P(X) \leftarrow \neg Q(X)$ .
- ... Fuzzy logic programs:  $p(X) \leftarrow 0.5 \cdot (q(x) + r(X))$ .
- ... Semi-ring constraint programs:  $p(X) \leftarrow 0.5 \otimes (q(x) \oplus r(X))$ .
- ? Probabilistic logic programs:  $0.3 :: p(X)$ .
- ? Hex-programs:  $tr(S, P, O) \leftarrow \&RDF[uri](S, P, O)$ .



## Operator-Based Semantics for other KR-formalisms

- autoepistemic logic [DMT03],
- default logic [DMT03],
- abstract argumentation [SW15],
- abstract dialectical frameworks [SW15],
- weighted abstract dialectical frameworks [Bog19],
- SCHACL [BJ21].

# Operator-Based Studies

Top-Down approach:

- Instead of studying a concept for a specific framework, define and study it for operators over a lattice (and their approximations).
- We can then apply this concept to all formalisms that are or can be captured in AFT.

Examples:

- |                                    |                                  |
|------------------------------------|----------------------------------|
| ✓ Stratification [VGD06]           | ✓ Argumentative dialogues [HA20] |
| ✓ Conditional Independence [Hey23] | ?                                |
| ✓ Knowledge Compilation [BVdB15]   | ? Belief dynamics                |
| ✓ Groundedness [BVdB15]            | ? Modular equivalence            |
| ✓ Strong equivalence [Tru06]       | ? Neuro-symbolism                |

## Round up

---

# Summary

- Operators as the core for understanding answer set semantics.
- Paved the road towards approximation fixpoint theory.
- Algebraic theory that allows language independent work on KR.
- Requires some buy-in, but in my view a great bargain.
- Interested in cooperating? Questions on AFT? Come talk to me.



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