



渕一博記念コロキウム
『論理と推論技術：四半世紀の展開』

非単調性と帰納論理を取り入れたことで
論理プログラミングはどう変わったか？

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Incorporating Nonmonotonic and Inductive Inference into Logic Programming

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Oliver Ray, Chiaki Sakama, Yoshitaka Yamamoto

History of Thinking Machines

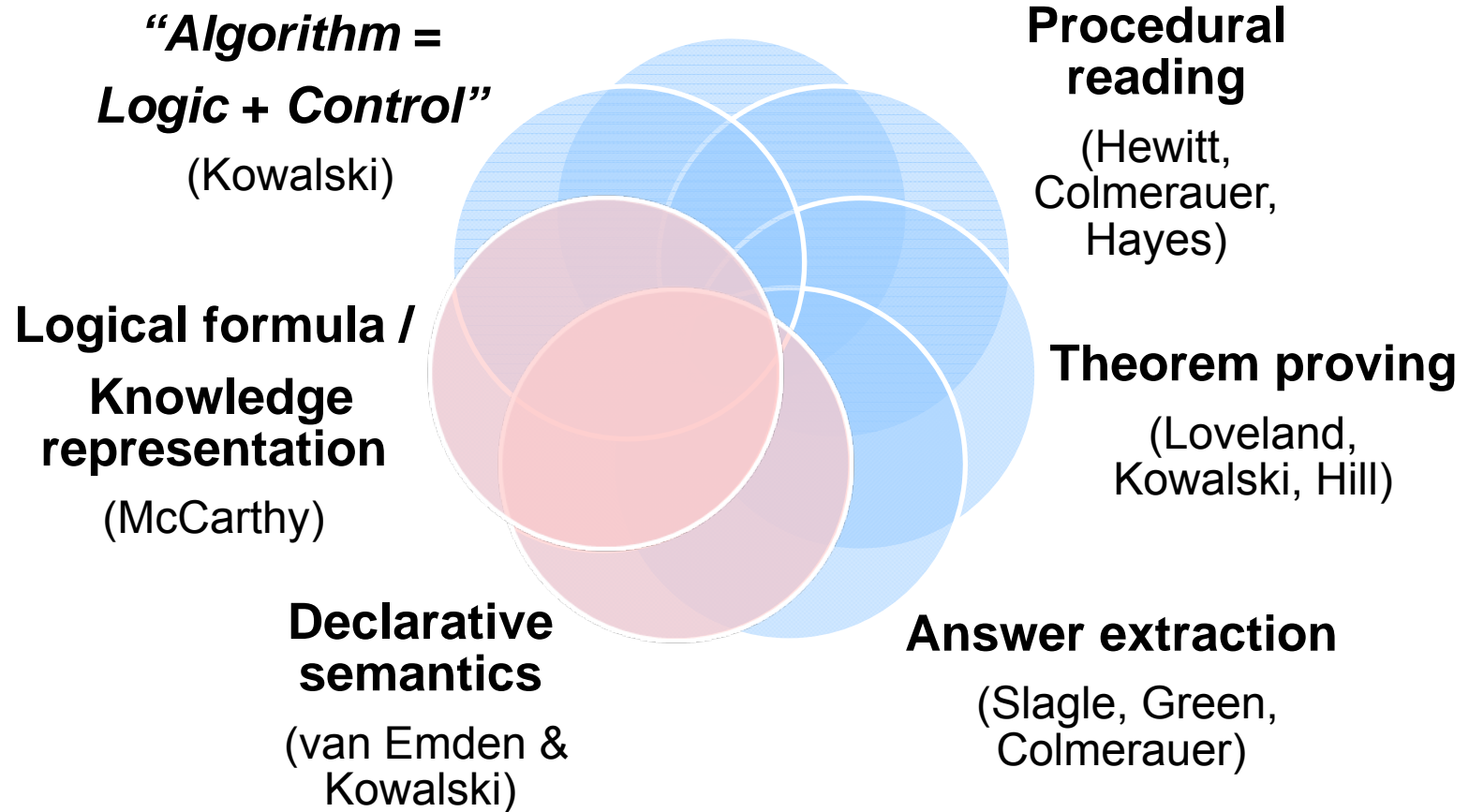
- 1936. Turing Machine
- 1943. McCulloch & Pitts: Boolean circuit model of brain
- 1950. Turing: "Computing Machinery and Intelligence"
- 1956. Dartmouth meeting: "Artificial Intelligence" adopted
- 1958. McCarthy: "Programs with Common Sense"
- 1950s. Early AI programs, including Samuel's checkers program, Newell & Simon's Logic Theorist
- 1960. McCarthy: LISP
- 1965. Robinson: complete algorithm for logical reasoning
- 1969. McCarthy & Hayes: situation calculus, frame problem
- 1971. Colmerauer & Kowalski: Prolog
- 1970s. Early development of knowledge-based systems
- 1980. McCarthy: circumscription
- 1982. **Japan's Fifth Generation Computer Project**
- 1986. Neural networks return to popularity
- 1995. Emergence of intelligent agents
- 1997. Deep Blue

History of Logic Programming

- 1960. Davis & Putnum: testing satisfiability of propositional formulas
- 1962. Davis, Logemann & Loveland: improvement of DP → SAT
- 1965. Robinson: resolution principle
- 1968. Loveland: Model Elimination
- 1971. Kowalski & Kuehner: SL-resolution
- 1972. Colmerauer & Roussel: “*Programmation en Logique*”
- 1972. Kowalski: “The predicate calculus as a programming language”
- 1973. Hill: LUSH (SLD-)resolution
- 1974. van Emden & Kowalski: Scott’s fixpoint & Tarskian semantics
- 1977. Warren (David): Edinburgh Prolog compiler
- 1978. Clark: negation as failure (predicate completion)
- 1982. **Japan's Fifth Generation Computer Project**
- 1980s. Shapiro, Clark, Ueda: Concurrent Logic Programming
- 1987. Jaffer & Lassez: Constraint Logic Programming
- 1988. Gelfond & Lifschitz: stable model semantics
- 1991. Muggleton: “Inductive Logic Programming”
- 1992. Kakas, Kowalski & Toni: “Abductive Logic Programming”
- 1990s. Poole, Sato: Probabilistic Logic Programming

Interpreting Horn Logic Programs

$$H \text{ :- } B_1, \dots, B_n.$$



Extending Horn Logic Programs

$H_1; \dots; H_m; \text{not } H_{m+1}; \dots; \text{not } H_n$

$\leftarrow B_1, \dots, B_m, \text{not } B_{m+1}, \dots, \text{not } B_n$

- Definite (Horn) program (H, B: atom)
- Normal logic program
- Extended logic program (H, B: literal)
- Extended disjunctive program
- General extended disjunctive program
- Nested program (H, B: rule)

Logic Programming and Nonmonotonic Reasoning

- 1969. McCarthy & Hayes: frame problem
- 1978. Reiter: closed-world assumption
- 1979. Clark: negation as failure (Compl)
- 1980. McCarthy: circumscription (Circ)
- 1980. Reiter: default logic (DL)
- 1982. Reiter: Circ implies Compl (sometimes)
- 1985. Moore: autoepistemic logic (AEL)
- 1987. Gelfond & Lifschitz: stratified LP as prioritized Circ
- 1988. Gelfond & Lifschitz: stable model semantics (LP as AEL)
- 1990. Gelfond & Lifschitz: answer set semantics (LP as DL)
- 1992. Inoue, Koshimura & Hasegawa: answer set computation in KL1
- 1996. Niemelä & Simons: smodels
- 1997. Eiter, Faber, Leone & Pfeifer: DLV
- 2007. Lifschitz, Lin: 2nd-order formalization of stable models (LP as Circ)

Answer Set Programming

A **program** is regarded as the **constraints** to be satisfied by **solutions**. Each solution is obtained by computing an **answer set (stable model)** of the program.

A program may have no, one, or multiple answer sets.

Program:

$p ; \mathit{not} p \leftarrow,$	}	generator
$q ; \mathit{not} q \leftarrow,$		
$r ; \mathit{not} r \leftarrow,$		
$s \leftarrow p, r,$	}	tester
$s \leftarrow p, q, \mathit{not} r,$		
$\leftarrow \mathit{not} s.$	}	goal

Answer Sets: $\{p, q, r, s\}, \{p, r, s\}, \{p, q, s\}.$

Answer Set Programming

A **program** is regarded as the **constraints** to be satisfied by **solutions**. Each solution is obtained by computing an **answer set (stable model)** of the program.

A program may have no, one, or multiple answer sets.

Program:

$p \leftarrow \textit{not } q, \textit{not } r,$	}	generator (exclusive choice)
$q \leftarrow \textit{not } p, \textit{not } r,$		
$r \leftarrow \textit{not } p, \textit{not } q,$		
$s \leftarrow p, r,$	}	tester
$s \leftarrow q, \textit{not } r,$		
$\leftarrow \textit{not } s.$	}	goal

Answer Sets: $\{q, s\}.$

Nonmonotonic Reasoning in ICOT

1984 Kitakami, Kunifuji, Miyachi & Furukawa: knowledge assimilation (SLP'84)

1986 Kunifuji, Tsurumaki & Furukawa: hypothesis-based reasoning

1986 Goebel, Poole & Furukawa: theory formation (ICLP'86)

1988 Ishizuka: *Hypothetical Reasoning WG*

1988 Arima, Satoh: work on circumscription (FGCS'88)

1989 Sakama: possible model semantics for disjunctive programs

1989 *Nicolas Helft & David Poole in ICOT*

1990 Inoue & Helft: circumscriptive theorem prover

1991 Helft, Inoue & Poole: QA in circumscription (IJCAI-91)

1991 *Mark Stickel in ICOT*

1991 Inoue: SOL-resolution (IJCAI-91)

1991 Inoue, Satoh & Iwayama: abduction in answer set semantics (ICLP'91)

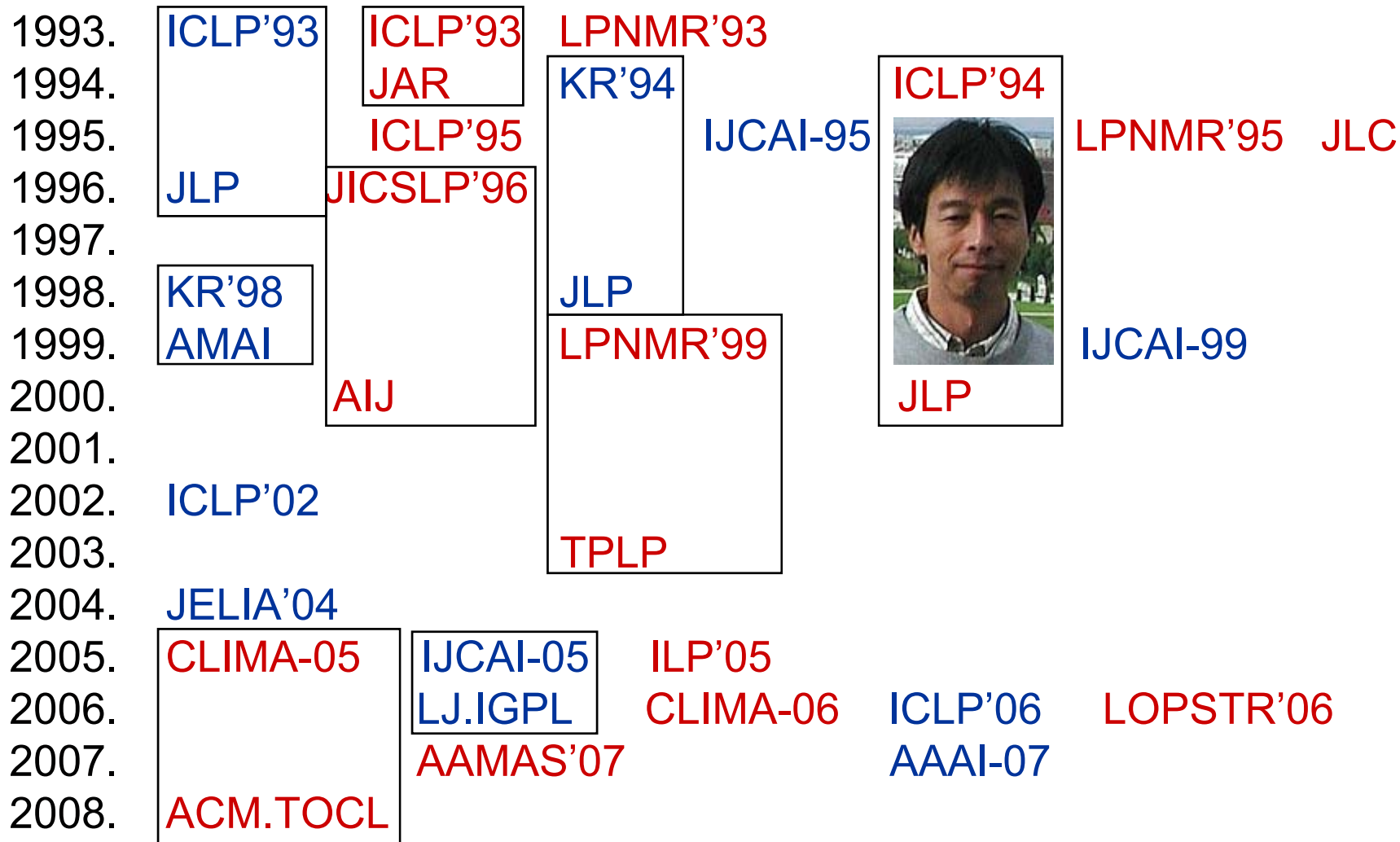
1992 Inoue, Koshimura & Hasegawa: answer set comp. on MGTP (CADE'92)

1992 Ohta & Inoue: parallel abductive reasoning on MGTP & ATMS (FGCS'92)

1993 Inoue, Ohta, Hasegawa & Nakashima: abduction on MGTP (IJCAI-93)

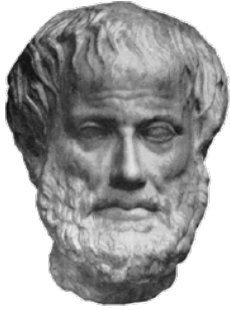
1993 Inoue & Sakama: fixpoint semantics for abductive programs (ICLP'93)

Inoue & Sakama or Sakama & Inoue



Abduction, Induction and Deduction

Analytic Reasoning



**Deduction
(LP)**

consequence:
from *prior knowledge*
to *necessary implications*

Synthetic Reasoning



**Induction
(ILP)**

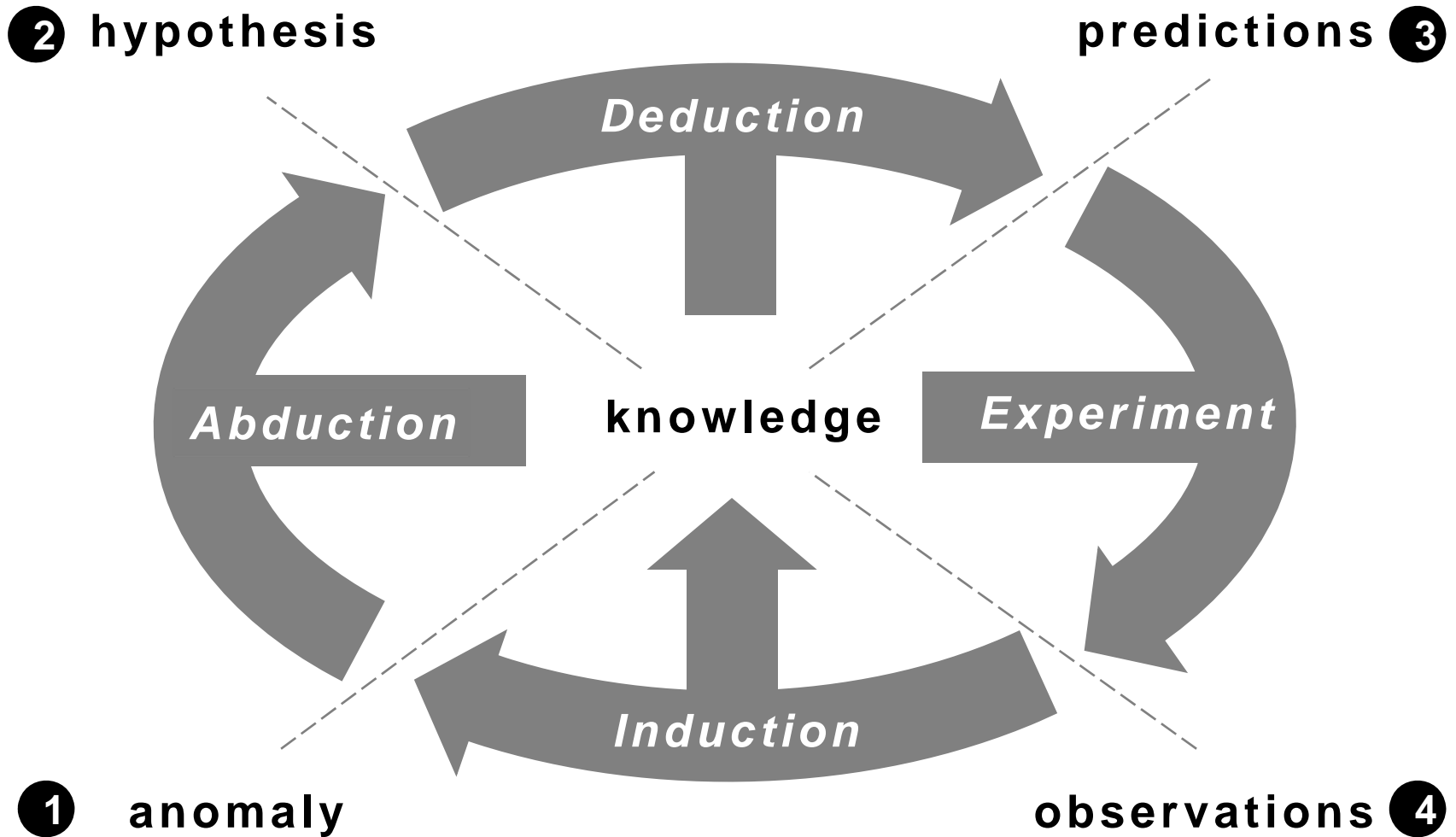
generalization:
from *observed samples*
to *wider populations*

**Abduction
(ALP)**

explanation:
from *given effects*
to *possible causes*

Scientific Discovery

Scientific Knowledge Development



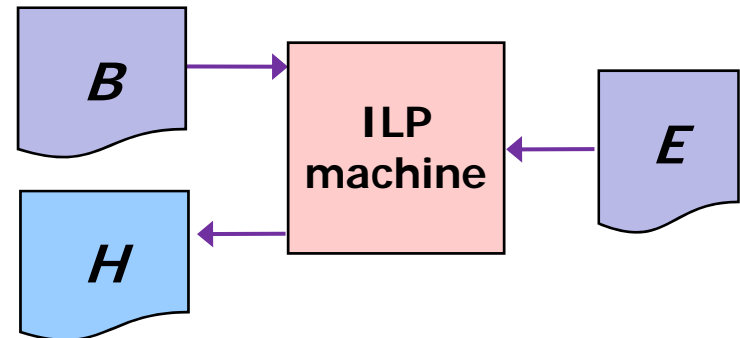
Abduction and Induction: Logical Framework

Input:

- B : background theory
- E : (positive) examples / observations

Output:

- H : hypothesis satisfying that
 - $B \wedge H \models E$
 - $B \wedge H$ is consistent.



Inverse Entailment (IE)

Computing a hypothesis H can be done **deductively** by:

$$B \wedge \neg E \models \neg H$$

We use a **consequence finding** technique for IE computation.

Consequence Finding

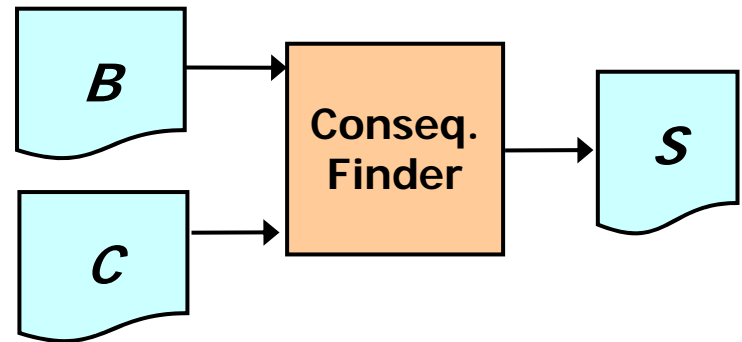
Input:

- B : first-order (clausal) theory
- C : “new” clausal theory
- P : language restriction (“*production field*”)

Output:

- S : the (minimal) “new” consequences satisfying that

- $B \wedge C \models S$
- $B \not\models S$
- S belongs to P .



- **SOL-resolution (Inoue, IJCAI-91)**
- **SOLAR (Nabeshima, Iwanuma & Inoue, TABLEAUX'03)**
- For Theorem Proving, C is the negation of the target theorem and S is the empty clause (generalization of *proof-finding*).
- For Inverse Entailment, $C = \neg E$ and $S = \neg H$: $B \wedge \neg E \models \neg H$.

Inverse Entailment for Abduction

SOLAR Example: graph completion problem – *pathway finding*

Find an arc which enables a path from A to D.

Background theory:

$\text{path}(X,Y) \leftarrow \text{node}(X), \text{node}(Y), \text{arc}(X,Y).$

$\text{path}(X,Z) \leftarrow \text{node}(X), \text{node}(Y), \text{node}(Z), \text{arc}(X,Y), \text{path}(Y,Z).$

$\text{node}(a). \text{node}(b). \text{node}(c). \text{node}(d). \text{arc}(a,b). \text{arc}(c,d).$

Negated observation:

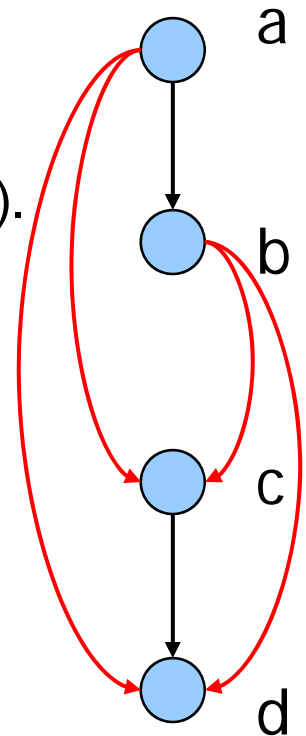
$\neg \text{path}(a,d).$

Production field:

literal form = $[\neg \text{arc}(_,_)]$ & clause length = 1.

Output of SOLAR:

1. $\neg \text{arc}(a, d).$ 2. $\neg \text{arc}(a, c).$ 3. $\neg \text{arc}(b, c).$ 4. $\neg \text{arc}(b, d).$



Metabolic Pathway (Ray & Inoue, DS'07)

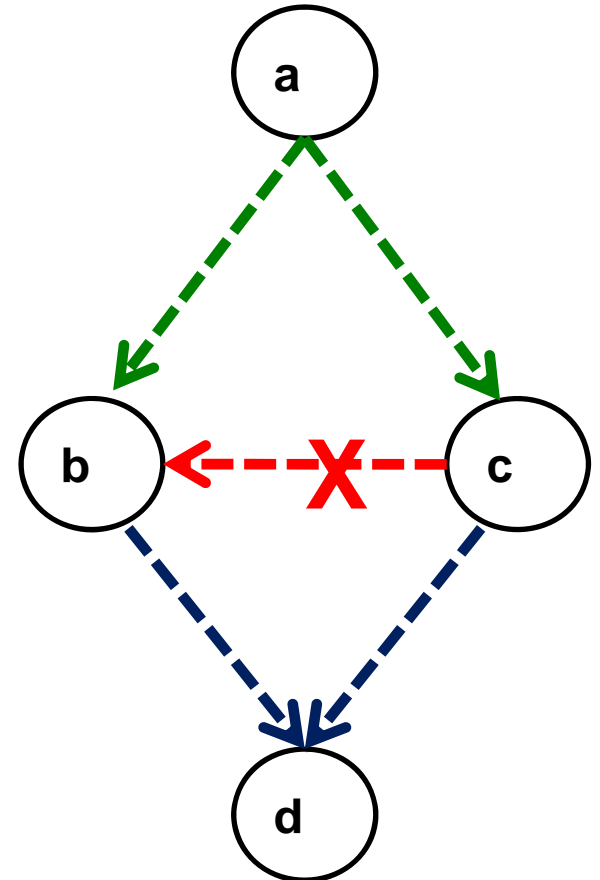
$$T = \begin{cases} \text{pathway}(X, Z) \leftarrow \text{reaction}(X, Y) \wedge \text{pathway}(Y, Z) \\ \text{pathway}(X, Z) \leftarrow \text{reaction}(X, Z) \\ \text{reaction}(a, b) \vee \text{reaction}(a, c) \\ \text{reaction}(b, d) \vee \text{reaction}(c, d) \\ \neg \text{reaction}(c, b) \end{cases}$$

$$E = \{ \text{pathway}(U, d) \}$$

% (from which U) is there a path to d?

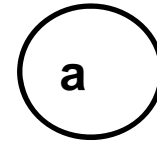
$$A = \{ \text{reaction}(V, W) \}$$

% assuming reactions from some V to W



Metabolic Pathway: A Solution

$$B = \begin{cases} \textit{pathway}(X, Z) \leftarrow \textit{reaction}(X, Y) \wedge \textit{pathway}(Y, Z) \\ \textit{pathway}(X, Z) \leftarrow \textit{reaction}(X, Z) \\ \textit{reaction}(a, b) \vee \textit{reaction}(a, c) \\ \textit{reaction}(b, d) \vee \textit{reaction}(c, d) \\ \neg \textit{reaction}(c, b) \end{cases}$$

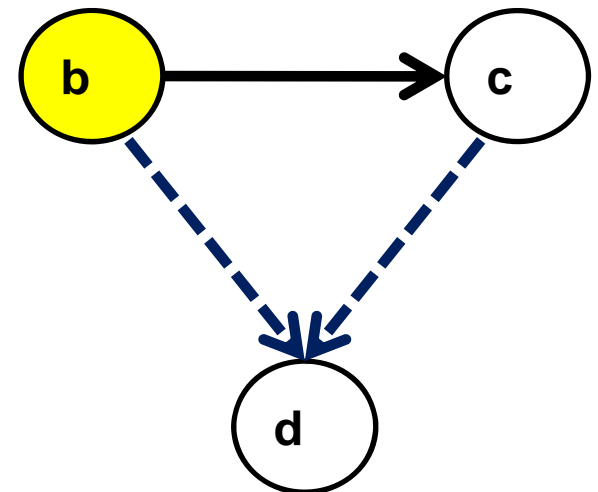


$$E\sigma = \{ \textit{pathway}(b, d) \}$$

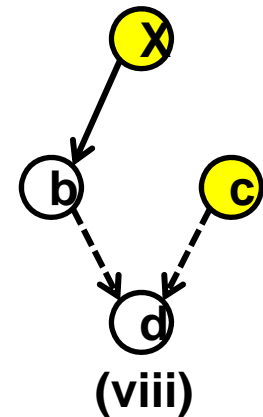
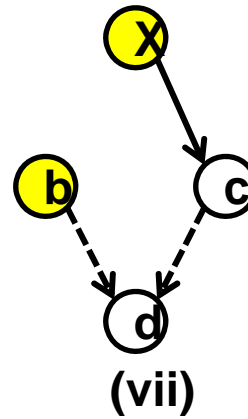
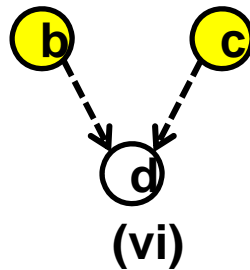
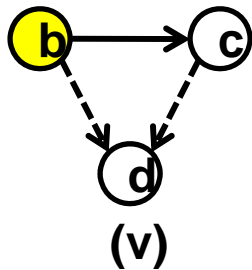
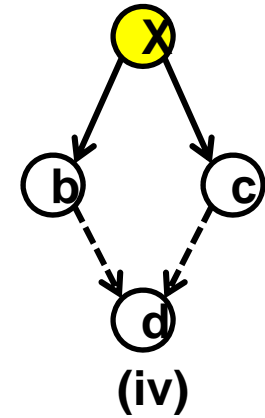
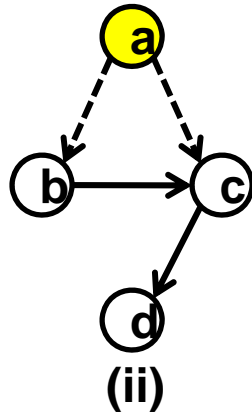
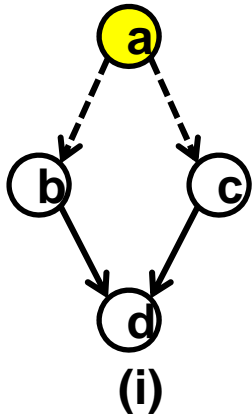
% there is a path from b to d

$$H = \{ \textit{reaction}(b, c) \}$$

% assuming a reaction from b to c



Metabolic Pathway: MORE Solutions



Problem: want to express non-ground answers like (iii), and disjunctive answers such as (vi).

Abductive Inference (Ray & Inoue, DS'07)

given

B theory (background knowledge)
 E goal (set of given observations)
 A abducibles (set of possible assumptions)

find

$H \subseteq A$ explanation (set of assumptions)
 Θ answer (SET of variable bindings)

where

$$B \models \forall \left(\bigwedge_{L \in H} L \rightarrow \bigvee_{\sigma \in \Theta} E \sigma \right)$$

% the conjunction of assumptions H implies the disjunction of answers Θ .

Metabolic Pathway: Another Solution

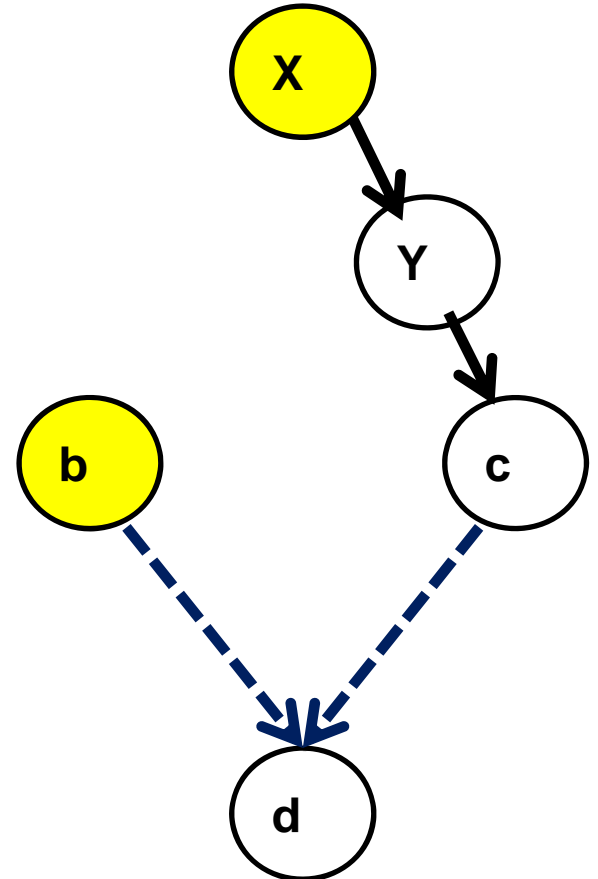
$$B = \begin{cases} \text{pathway}(X, Z) \leftarrow \text{reaction}(X, Y) \wedge \text{pathway}(Y, Z) \\ \text{pathway}(X, Z) \leftarrow \text{reaction}(X, Z) \\ \text{reaction}(a, b) \vee \text{reaction}(a, c) \\ \text{reaction}(b, d) \vee \text{reaction}(c, d) \\ \neg \text{reaction}(c, b) \end{cases}$$

$$\Theta = \{ \{U / b\}, \{U / X\} \}$$

% there is a path from b or X to d

$$H = \{ \text{reaction}(X, Y), \text{reaction}(Y, c) \}$$

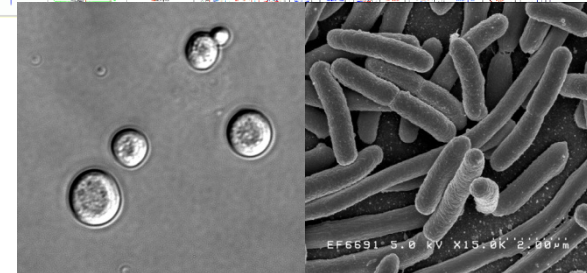
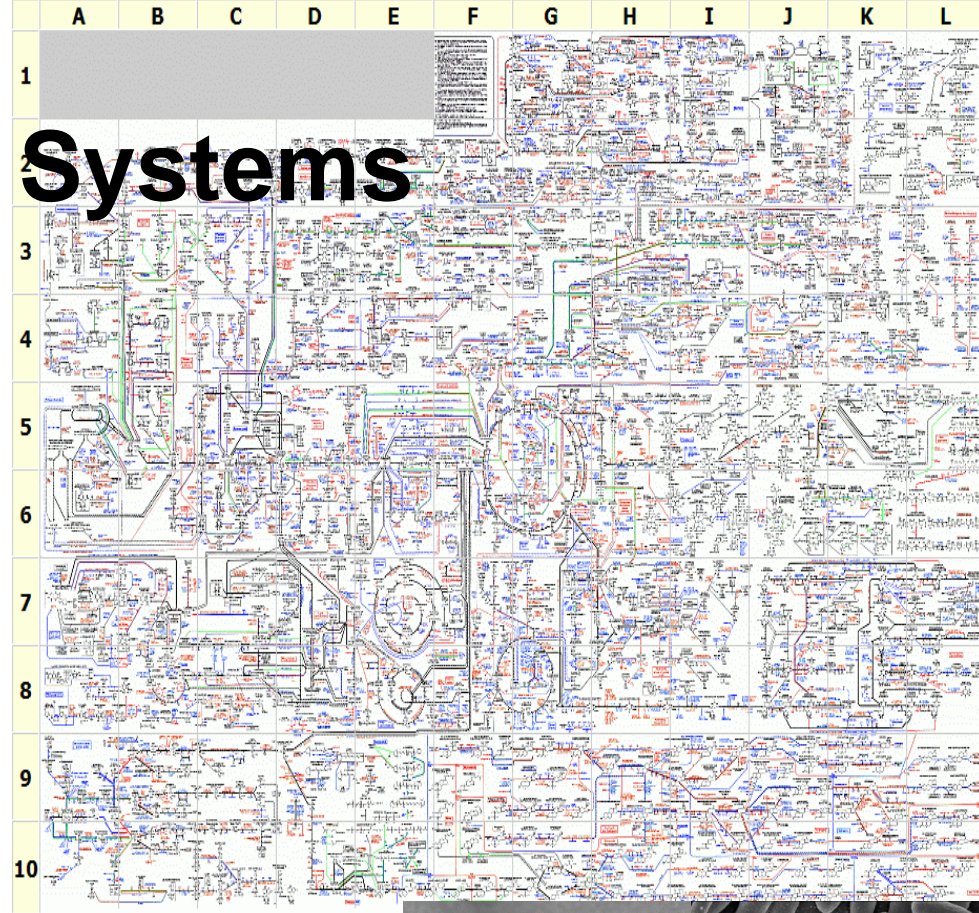
% assuming reactions from X to Y and from Y to c



Modeling Biological Systems

- Explain and predict metabolic pathways.
 - Generic Model:
 - *Saccharomyces Cerevisiae*
 - *E-coli*

– Biological Phenomenon can be explained by *Inductive Logic Programming (ILP)*.



Inductive Learning Approaches

■ Goals

- Finding inhibitions in a metabolic pathway.
- Discovering causal rules which augment an incomplete background theory.
- Predicting changes of concentration in intracellular fluxes.

■ Previous Work

- Using an **abductive logic programming** technique on the problem of inhibitions of metabolic pathways at steady states (Tamaddoni-Nezdah et al., 2006)

■ **New Approach** (Yamamoto, Inoue & Doncescu, 2007)

- Integration of **abduction** and **induction**.
- Not only **steady states** but also **dynamic models**.

Inverse Entailment for Induction

- **Horn clauses** for background theories
 - **Progol** (Muggleton, 1995)
- **Full clausal theories** (non-Horn clauses) for background, example, and hypothesis theories.
 - **CF-induction** (Inoue, ILP-2001; Yamamoto, Ray & Inoue, LLLL-2007)
 - **fc-HAIL** (Ray & Inoue, ILP-2007)
- **Note:** CF-induction is the only existing ILP system that is **complete** for full clausal theories.

Example: Metabolic Pathway (Pyruvate)

■ *B*:

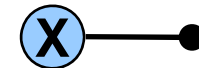
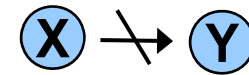
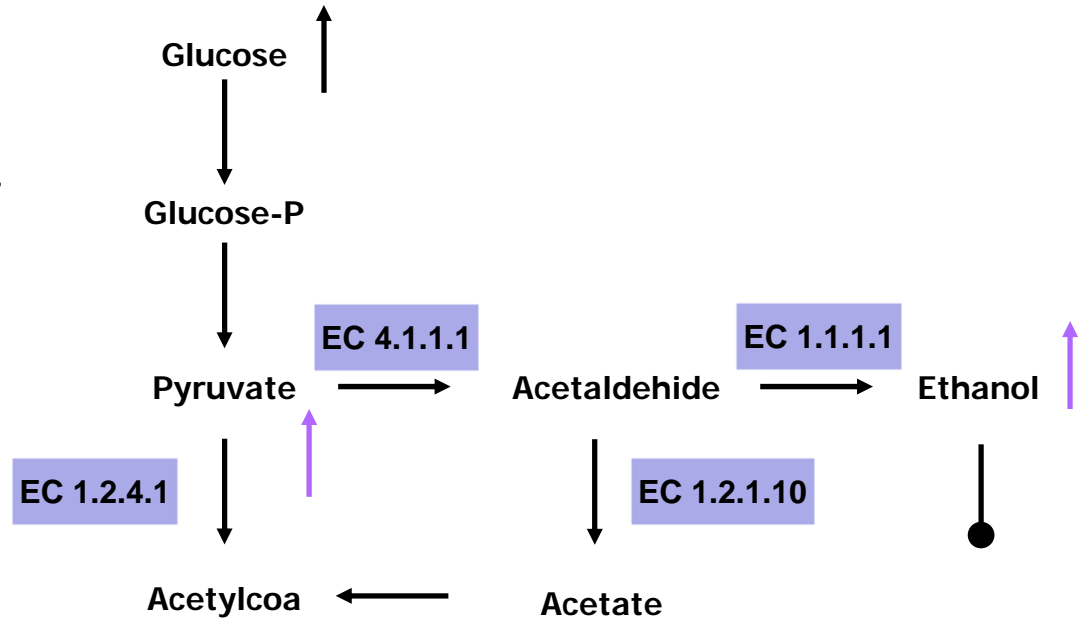
reaction(pyruvate, acetylcoa).
 reaction(pyruvate, acetaldehyde).
 reaction(glucose, glucosep).
 reaction(glucosep, pyruvate).
 reaction(acetaldehyde, acetate).
 reaction(acetate, acetylcoa).
 reaction(acetaldehyde, ethanol).
 concentration(glucose, up).
 terminal(ethanol).

blocked(X) ← reaction(X,Y), inhibited(X,Y).

blocked(X) ← terminal(X).

concentration(X,up) ← reaction(Y,X), ¬inhibited(Y,X), blocked(X).

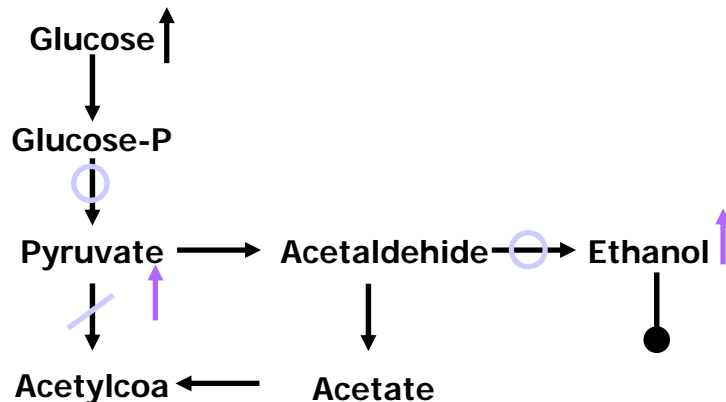
■ *E*: concentration(ethanol,up). concentration(pyruvate, up).



Example: Outputs of CF-induction

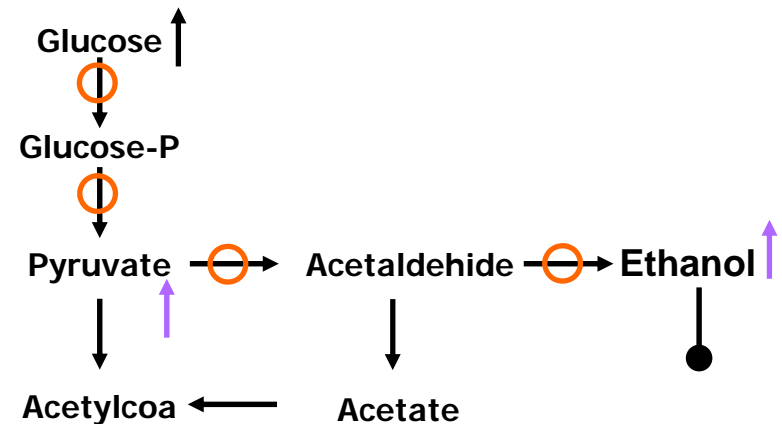
■ H_1 :

- ¬Inhibited(glucosep, pyruvate).
- ¬inhibited(acetaldehyde, ethanol).
- inhibited(pyruvate, acetylcoa).



■ H_2 :

- ¬inhibited(glucose, glucosep)
- ¬Inhibited(glucosep, pyruvate).
- ¬inhibited(acetaldehyde, ethanol).
- ¬inhibited(pyruvate, acetaldehyde).
- concentration(Y, up) ←
- ¬inhibited(X, Y), concentration(X, up).



Conclusion

- Algorithm = Logic + Control.
- Logic programming is still alive.
- Is this GOFAI?
- Dealing with **incomplete knowledge** and **elaboration tolerance** — default reasoning, planning, prediction, explanation, hypothesis formation, model generation.
- Inference-based hypothesis-finding in biochemical networks.
- Logic is a glue to combine mathematical/abstract models with the real world.
- Get back to logic (programming)!