

Chapter 11. Knowledge Representation and Reasoning

The Quest for Artificial Intelligence, Nilsson, N. J., 2009.

Lecture Notes on Artificial Intelligence

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Overview

- Methods for knowledge representation and reasoning from Mid-1960s and Mid-1970s
 - Symbolic logic and its deductions
 - Predicate calculus
 - For proving theories
 - Situation calculus
 - Logic programming: PROLOG
 - Sematic networks: HAM, MEMS, MENTAL
 - Script and Frames

Introduction

■ Knowledge

- For intelligent system
- The mean to draw conclusion from or act on

■ Knowledge representation

- Procedural
 - Coordinate and control the specific action (ex. hitting a tennis ball)
 - Programs using the knowledge
 - Specific task program
- Declarative
 - Declarative sentence (I am a 25 years old)
 - Symbolic structures
 - General task program

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11.1 Deductions in Symbolic Logic

Deductions in Symbolic Logic

■ The predicate calculus

- From Aristotle to G. Boole and McCarthy

- Ex. Aristotle syllogism

- 1. $(\forall x)[\text{Man}(x) \supset \text{Mortal}(x)]$

(The expression “ $(\forall x)$ ” is a way of writing “for all x”; and the expression “ \supset ” is a way of writing “implies that.” “ $\text{Man}(x)$ ” is a way of writing “x is a man”; and “ $\text{Mortal}(x)$ ” is a way of writing “x is mortal.” Thus, the entire expression is a way of writing “for all x, x is a man implies that x is mortal” or, equivalently, “all men are mortal.”)

- 2. $\text{Man}(\text{Socrates})$ (Socrates is a man.)

- 3. Therefore, $\text{Mortal}(\text{Socrates})$ (Socrates is mortal.)

- “Therefore,” is an example of a *deduction*

- Rules of inference (ex. Modus ponens)

Deductions in Symbolic Logic

- Early works on deduction in symbolic logic
 - Programs using inference rule (1960s) for proving theorems in the predicate calculus
 - P. Gilmore, H. Wang, and D. Prawitz (IBM)
 - F. Black (Harvard)
 - QA3 (Question Answering)
 - C. C. Green implemented a new deduction method developed by J. A. Robinson
 - From two other statements, a new statement is generated by rules (ex. $P \vee \neg Q$ and P produces Q)
 - Key contribution: how resolution could be applied to general expressions in the predicate calculus
 - Example
- So just as with programs for playing games, LT, and proving geometry theorems, deduction programs need to try many possibilities in their search for a solution

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11.2 The Situation Calculus

The Situation Calculus

■ Situation calculus

- Where one could write logical statements that explicitly named the situation in which something or other was true
- Ex. “What is a program for rearranging a list of numbers so that they are in increasing numerical order?”
- Block case
 - block A is on top of block B in some situation S
→ $\text{On}(A, B, S)$
 - block A is blue in all situations
→ $(\forall s)\text{Blue}(A, s)$
 - there exists some situation in which block A is on block B
→ $(\exists s)\text{On}(A, B, s)$
- QA3 can deduce situation calculus → robot plan

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11.3 Logic Programming

Logic Programming

- Green's automatic programming
 - QA3 can construct simple computer programs
 - The first attempt to write programs using logical statements
- SL-resolution : A. Kowalski and D. Kuehner
- PROLOG (1972)
 - A. Comerauer, P. Roussel, and A. Kowalski
 - An ordered sequence of logical statements
 - The exact order in which these statements are written, along with some other constructs, is the key to efficient program execution

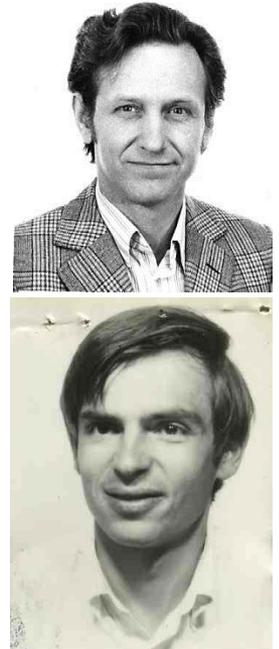


Figure 11.1: Robert Kowalski (top) and Alain Colmerauer (bottom)

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11.4 Semantic Networks

Semantic Networks

- **Semantic networks**
 - Another format for representing declarative knowledge
- **Human Associative Memory (HAM)**
 - G. Bower and J. Anderson (1970s)
 - Network-based human memory
 - Parse simple propositional sentences and store them in the semantic network structure
 - With accumulated memory, HAM can answer simple questions
- **MEMS and MENTAL: S. C. Shapiro (1971)**
 - MEMS: a network structure for storing semantic information
 - MENTAL: aided MEMS in deducing new information from that already stored
- **SNePS: S. C. Shapiro**
 - Combination of logical representation with those of network representations used for natural language understanding

Semantic Networks

- Conceptual dependency representations for natural language sentences
 - R. C. Schank
 - People transform natural language sentences into “conceptual structures independent of the particular language where the sentences were expressed.

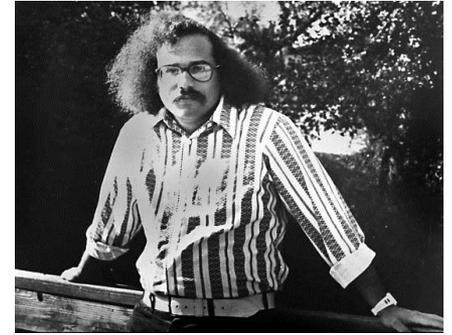


Figure 11.2: Roger Schank.

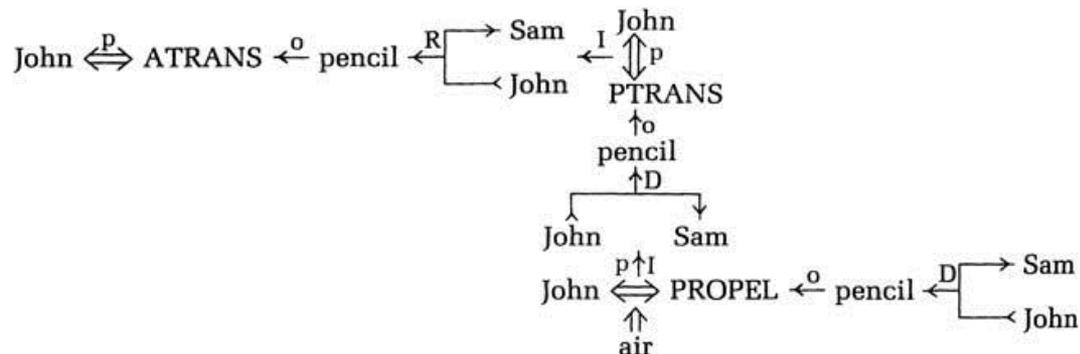


Figure 11.3: Conceptual structure for "John threw the pencil to Sam." (From Roger C. Schank, "Identification of Conceptualizations Underlying Natural Language," in Roger Schank and Kenneth Colby (eds.), *Computer Models of Thought and Language*, p. 226, San Francisco: W. H. Freeman and Co., 1973.)

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11.5 Scripts and Frames

Scripts and Frames

■ Graphical knowledge representations

- Semantic networks and conceptual structures
- Efficient computationally due to participating in the same chain of reasoning

■ Scripts

- Proposed by R. Schank and R. Abelson
- A script is a way of representing what they call “specific knowledge – detailed knowledge about a situation or event that “we have been through many times.”
- Example

■ Frames

- Proposed by M. Minsky
- a data-structure for representing a stereotyped situation, like being in a certain kind of living room, or going to a child's birthday party.
- Implementation: FRL and KRL

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Appendix

Deductions in Symbolic Logic

- QA3
 - Resolution-based deduction system
 - The advantage of resolution
 - Implemented in programs to make deductions from a set of logical statements consisting of “clauses”
 - Ex.
 - 1. ROBOT(Rob) (Rob is a robot.)
 - 2. $(\forall x)[\text{MACHINE}(x) \supset \neg \text{ANIMAL}(x)]$
(x is a machine implies that it is not an animal.)
The system is then asked “Is everything an animal?” by having it attempt to deduce the statement
 - 3. $(\forall x)\text{ANIMAL}(x)$
QA3 answers “NO” and gives a “counterexample”
 - 4. $x = \text{Rob}$
(This indicates that $\text{ANIMAL}(\text{Rob})$ contradicts what was to be deduced.)

Scripts and Frames



■ An example of scripts

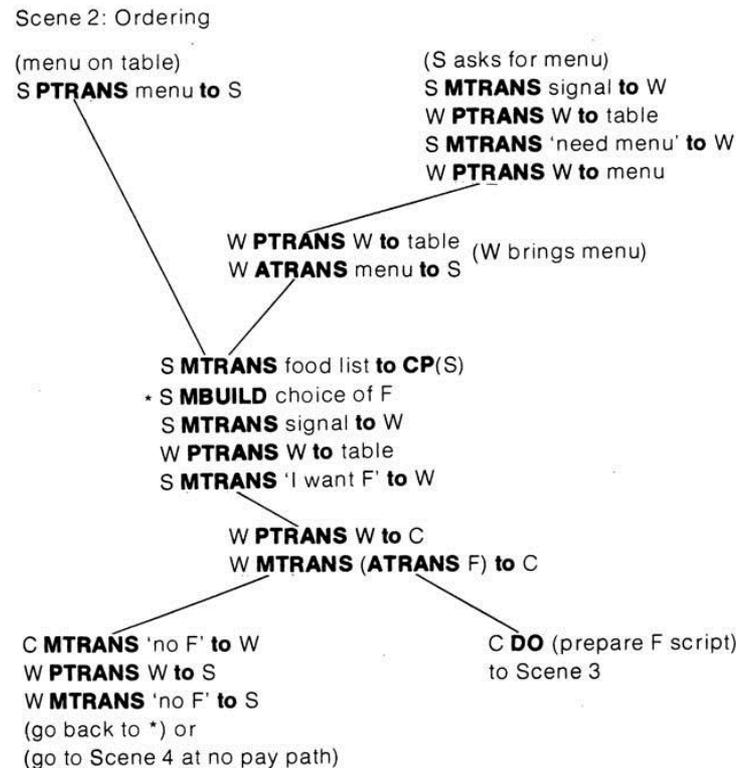


Figure 11.4: A scene in the restaurant script. (From Roger C. Schank and Robert P. Abelson, *Scripts, Plans, Goals, and Understanding: An Inquiry into Human Knowledge Structures*, p. 43, Hillsdale, NJ: Lawrence Erlbaum Associates, 1977.)