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#### Distributed Communicating Agents: Guide for Installation and Operation

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#### Abstract

This report describes a software tool for developing distributed, knowledge-based, communicating agents. The tool can be used to construct individual expert systems that communicate and cooperate with each other, and with human agents, in solving problems. The agents interact by using Rosette, an actor-based language, to manage their communications through TCP/IP and OSI. Thus, they can be located anywhere that is reachable through OSI. The agents also use Rosette to access databases. You can use this guide to construct, install, operate, and interact with teams of agents.

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		8
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# Contents

1	Intr	Introduction to Distributed Agents												
	1.1	Major	Features of RAD	3										
2	Dis	tribute	ed Reasoning	8										
	2.1	Model	for Distributed Reasoning	9										
	2.2		ecture and Use											
		2.2.1		13										
		2.2.2		15										
		2.2.3	Remote Justifications and Distributed TMS											
			Negotiation											
	2.3			17										
	2.4		mentation Issues	22										
			Protocol for Message Passing											
			Extensible Services Switch (ESS)											
3	Tru	ıth Ma	intenance	27										
4	Fra	mes ar	nd Assertions	32										
	4.1													
	4.2													
	4.3		oles and Types											
	4.4		ions											
		4.4.1		<b>3</b> 9										
		4.4.2		41										
5	Rel	lations	in RAD	44										
_				44										

	5.2	Relations on Relations
	5.3	Predefined Relations
		5.3.1 Type Relations
		5.3.2 Comparison Relations 51
		5.3.3 Metarelations
		5.3.4 Pathological Relations 54
		5.3.5 RAD Knowledge Base Relations 54
		5.3.6 Class System Relations
		5.3.7 Interface with Lisp
		5.3.8 Querying the User
		5.3.9 Miscellaneous Built-In Relations
	5.4	User-Defined Relations
		and the second s
6	Bac	kward Inference 61
	6.1	Asserting Proof Results
	6.2	Proof Strategies
		6.2.1 Relations
		6.2.2 Multiple-Valued Relations
		6.2.3 Single-valued Relations 67
7	For	ward Inference 68
*	7.1	Overview of Forward Chaining
	7.2	Forward Chaining in More Detail
	1.2	Forward Chaining in More Detail
8	Cor	ntradiction Resolution 74
	8.1	Introduction
	8.2	The Contradiction Resolution Process
		8.2.1 The FIX Phase
		8.2.2 Dependency-Directed Backtracking 81
	8.3	User Hints
		8.3.1 General Assertions
		8.3.2 Retractable Default Assumptions 85
		8.3.3 Preferences
	0 1	Final Cayotta 88

9	Synt		the User Interface													90
	9.1		yntax													
	9.2	The $R$	AD Reader												•	96
		9.2.1	The Slash $-/\ldots$	*										•		97
		9.2.2	RAD Reader Macros .													97
		9.2.3	Error Detection													
	9.3		vel Commands													
		9.3.1	Creating and Modifyin													
		9.3.2	Querying and Examinia													
		9.3.3	Customizing the Interf													
		9.3.4	Miscellaneous Comma													
10	Inst	allatio	1													118
	10.1	Enviro	nment													118
			3													
			AKE Process													
Δ	Mul	tiagen	Tic-Tac-Toe													121
			edge Base for Referee								 					121
			edge Base for Player A													
	A 3	Knowl	edge Base for Player B								 					136
			g a Game													
B	Res	alving	Contradictions													144
ם	B.1	Simple	Example of CRM											_		144
		Comp	ex Example of CRM.	• •	•	•	•	•	•	•		Ĭ	1000			147
	10.4		Knowledge Base for E													
			Protocol for Use of CF													
		B.2.2	Protocol for Use of Cr	IMI												140

			U
×			
			Ī
			Û
			0
			LI Fi
			Ų

# Chapter 1

# Introduction to Distributed Agents

Knowledge-based systems have become an important part of computing. After more than a decade of successful exploitation, there are now over 100,000 fielded systems [Feigenbaum 1988]. Most of these systems are small, involving small knowledge bases, single reasoning strategies, and specific domains of application. However, several trends have recently become apparent:

- Applications requiring an AI approach are increasingly large and diverse, yet existing expert-system tools work well only for small, specialized problems. Larger problems cannot be solved either because the necessary systems are too complex to build and maintain or because no single approach is adequate for all aspects of the problem. New tools are required that allow modular development of knowledge bases.
- Information resources abound throughout our environment, but there is no way for a single reasoning system to exploit the information available. Reasoning systems must be able to communicate and cooperate with one another and access multiple sources of information to solve a problem, just as human experts do.
- Once an expert system is developed with today's shells, it becomes an isolated island of knowledge. It cannot easily be reused in other, similar problem domains. Because of this, the cost of engineering and maintaining multiple expert systems is already becoming noticeably

high in the business community, where AI applications are increasingly commonplace.

• Knowledge based systems share a set of common needs, both for representing knowledge and for reasoning with it. The cost of building a new system in a new domain can be substantially reduced if it is written in a high-level language (often called a shell) that satisfies these common needs with built-in mechanisms.

These trends suggest the need for a new kind of system building tool, a tool that supports the construction of distributed, communicating, knowledge-

based agents.

Distributed reasoning enables a set of knowledge-based systems, constructed quasi-independently, to act as a set of cooperating agents, working together to solve a problem. Developers of distributed reasoning systems can exploit a divide-and-conquer approach to development; they will be able to build smaller, more manageable knowledge-based agents. Smaller agents may represent alternative points of view on a problem. These smaller agents can also be reused in different combinations for solving additional problems as they arise. A further advantage of this approach is that it enables systems to be physically distributed in the world, just as the problems that they address are.

In this report we describe a tool, called RAD, that makes it possible to build distributed systems that solve large, complex reasoning problems. RAD is an extension of earlier MCC-proprietary expert system software. It provides high-performance forward and backward reasoning using Warren Abstract Machine (WAM) technology [Warren 1983], a frame system integrated into a typed unification algorithm [Aït-kaci, et al. 1985], a justification-based truth maintenance system [Russinoff 1985], and a contradiction resolution mechanism [Petrie 1987]. Recent extensions support OSI communications through Rosette [Tomlinson et al. 1991], access to multiple databases, and a substantially improved user interface.

RAD is a product of an ongoing research effort within the Carnot Project, whose goal is to provide tools for accessing and maintaining the consistency of data in distributed heterogeneous environments. This report describes how to install, construct, operate, and interact with multiple computational agents. Later reports will detail important applications and uses of these

agents for intelligent X.500 directory service, robust negotiation-based EDI, intelligent information retrieval, administration of multiple heterogeneous databases, and management of data consistency. We intend to exploit this version of RAD in building representative applications, in cooperation with our shareholders, and we intend to release a series of new RAD platforms that incorporate the results of those efforts.

A word of warning is appropriate here. Although RAD is intended to be a widely useful tool, the current release is only a research prototype: it is not a production-quality tool. Its only users so far are members of our research group. So if you decide to try to use it, please get in touch with us so that we can help. We are knowledgeable in several application areas, and will gladly give advice on the use of its novel features. We also value user input and guidance into the ongoing research effort.

# 1.1 Major Features of RAD

The other chapters of this report describe in detail each of RAD's components. The rest of this section provides the context for those descriptions by quickly surveying the major components of the system and pointing out where the corresponding detailed descriptions can be found.

#### Knowledge Representation

Early artificial intelligence systems relied on first-order predicate logic as a language for representing domain knowledge. While this scheme is completely general and semantically clear, it has been found to be inadequate for organizing large knowledge bases and encoding complex objects. As an alternative, various frame-based languages have been employed. These languages are designed to support the natural representation of structured objects and taxonomies. They have proved to be well-suited for representing many useful relations, although they lack the general expressive power of the predicate calculus.

RAD combines the best features of these alternatives. It is frame-based, but allows knowledge to be expressed in terms of arbitrary predicates. It also represents relations as frames, so that a relation can have properties, any of its arguments can be typed (which RAD will then enforce), it can

have arbitrary arity, and it can inherit information from other frames. The representation of relations in RAD is described in Chapter 5.

In Chapter 4, we introduce frames, along with classes, attributes, metaclasses, and types. This chapter also describes simple data, including assertions of arbitrary arity that are attached to frames. Here we discuss the use of a justification-based truth maintenance system in connection with singlevalued relations and inheritance.

#### Inference Mechanisms

Knowledge-based systems may also be classified according to inference methods. Most deductive systems may be characterized as either goal-directed (backward chaining) or data-directed (forward chaining). In a goal-directed system, logical implications are encoded as rules that are used by the system to reduce goals to simpler subgoals. This allows knowledge to be represented implicitly, without using space in the knowledge base, until it becomes relevant to a current problem. In this framework, however, it is difficult for the knowledge base designer to build control into a system. Data-directed inference, on the other hand, is based on production rules, which the system uses to derive all logical consequences of new data automatically. While control of inference is more natural within this paradigm, it uses space less efficiently, representing all knowledge explicitly.

RAD integrates both goal-directed and data-directed inference, allowing a knowledge engineer the freedom to decide whether each logical implication is more suitably represented as a backward rule or a forward rule. It also allows many inferences to be made quickly and efficiently through inheritance in the frame structure. Chapters 6 and 7 describe the RAD backward and forward chaining inference systems, respectively, and their integration with a truth maintenance system.

#### Truth Maintenance System

Another central feature of RAD is a nonmonotonic justification-based truth maintenance system (JTMS), which is integrated with the inference system at the architectural level. Based on [Doyle 1979], the RAD JTMS records logical inferences and dependencies among data as they are derived by both forward and backward chaining. This allows efficient revision of a set of

beliefs to accommodate new information, the retraction of a premise, or the discovery of a contradiction [Petrie 1989]. It also facilitates the generation of coherent explanations, which can be used for explanation-based learning as demonstrated by the MCC Argo project [Huhns and Acosta 1988]. We will see in Chapter 2 how it also forms the basis for negotiation among cooperating agents [Huhns and Bridgeland 1991]. Chapter 3 contains a discussion of data dependencies and the RAD JTMS. Chapter 8 describes the contradiction resolution mechanism.

#### Distributed Reasoning

The distributed reasoning facility allows reasoning to be distributed across multiple symbolic reasoners that can thus collaborate in solving a problem. To do this requires four things:

- A protocol for interagent communication,
- An "interpreter" to make the agents interpretable to each other in the case that they were developed using different representational frameworks,
- A common knowledge base of problem solving strategies that serves as the basis of collaboration, and
- A generalization of key reasoning mechanisms that transforms them from global methods suitable for use by a single agent into semilocal methods that can be distributed across agents.

The current release of RAD provides a protocol for interagent communication as well as a generalization of the JTMS facility. It allows multiple agents, each implemented in RAD, to communicate with each other. It supports a distributed truth maintenance system (DTMS) so that each agent can rely on the results of another's reasoning without having to keep track of the details of that reasoning. This DTMS enforces local consistency within each agent, while enabling negotiation about inconsistencies among agents.

Future versions will include communication protocols that will enable agents constructed in other rule-based languages, such as CYC and OPS5, to interact with RAD agents. Future versions also will increase the effectiveness and efficiency of the RAD agents by providing a common knowledge

base of problem solving methods. This knowledge base will support models for the beliefs, goals, and intentions of each agent. Agents will then have an understanding of each other and the roles that they play in an overall application. Their actions will then be flexible, but robust, and applicable to dynamic real-time problems such as those that arise when acccessing and managing integrated networks of information.

#### **High-Performance Implementation**

RAD represents a refinement of the Proteus language coupled with a high performance implementation based on the Warren Abstract Prolog Machine (WAM) [Warren 1983], which provides inference speed comparable to that of the fastest commercial Prolog systems. Instructions are generated from RAD rules by a WAM compiler and interpreted by a WAM emulator, both written in Lisp. One advantage of this scheme is that it facilitates the extension of the WAM instruction set to provide the functionality necessary for expert system applications, allowing the integration of the WAM with other Lisp-based RAD components [Bridgeland 1989]. Also, since the WAM instructions are interpreted directly by RAD (in contrast to Prolog systems that compile them further into native machine code), RAD programs generate relatively compact code, thereby allowing larger applications.

#### User Interface

Chapter 9 is devoted to facilities designed to aid the system builder. This chapter discusses the user-interface processes, the RAD built-in relations and commands, the syntax for communicating with agents, and the parser that interprets this syntax.

The user interface in *RAD* is implemented as a pair of processes, one to monitor a keyboard for input from a user and the other to communicate with computational agents. Eventually, it will allow communication with computational agents constructed in other expert system languages, such as OPS5 and KEE.

#### Current Implementation of RAD

RAD is implemented in a combination of Common Lisp, Rosette, and C. The main representation and symbolic reasoning mechanisms are implemented in

Common Lisp. The user interface and the communication component of the distributed reasoning system are implemented in C. Future releases may contain a substantially larger proportion in C, because we are exploring the possibility of implementing all of RAD in either C or C++. The network component of RAD is written in Rosette to take advantage of the OSI and TCP/IP communication protocols it provides. A window interface is implemented using X Windows.

A number of expert system applications developed by MCC share-holders [Steele 1989, Kodak 1988, Kirchen 1989, Harp and Sederberg 1988, Virdhagriswaran et al. 1987, Virdhagriswaran and Pitts 1987], based on Proteus, have confirmed the viability of our approach and the utility of the RAD functionality, particularly for the solution of design problems. The Appendix describes several additional examples of RAD usage.

# Chapter 2

# Distributed Reasoning

RAD supports two kinds of agents—human and computational—as well as databases, as shown in Figure 2.1. It does this by making use of recent technological advances in communication protocols and distributed artificial intelligence (DAI). DAI is concerned with the cooperative solution of problems by a decentralized group of agents. It is the appropriate technology for applications where

- expertise is distributed, as in design;
- information is distributed, as in office automation and enterprise integration;
- data are distributed, as in distributed sensing;
- decisions are distributed, as in manufacturing control; and
- knowledge bases are developed independently but must be interconnected or reused, as in next-generation knowledge engineering.

This chapter describes the primitives and features that *RAD* incorporates to support distributed reasoning among both human and knowledge-based agents. The agents are then able to cooperate in solving problems, to share expertise, to work in parallel on common problems, to be developed and implemented modularly, to be fault tolerant through redundancy, to represent multiple viewpoints and the knowledge of multiple human experts, and to be reusable. Additional benefits are discussed in [Huhns 1987], [Bond and Gasser 1988], and [Gasser and Huhns 1989].

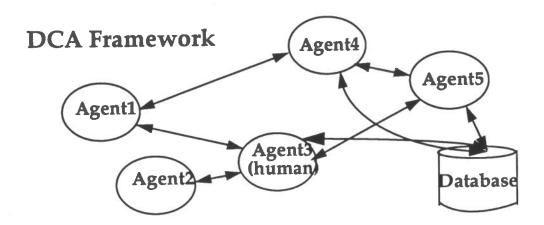


Figure 2.1: Architectural overview of the distributed communicating agent environment

## 2.1 Model for Distributed Reasoning

The following is a typical scenario for the use of RAD: there is a loosely-coupled network of experts, each being an expert in a particular field, and there is a problem that these experts must solve that is beyond the capabilities of any one of the experts. These experts can together solve the problem, but they must cooperate to do so. Furthermore, some of these experts might be knowledge-based systems, i.e., computational agents, some might be humans, and the remainder might be lower-level computational entities, such as databases, software simulators, and neural nets. (In this chapter, we refer generically to all of these as "agents.") RAD provides specific assistance for the development of computational agents, provides interfaces for humans to interact with other agents, and provides the overall framework within which all of these kinds of agents can operate and interact. Their interaction enables their cooperation and, ultimately, the solution of their problem.

The RAD framework and its corresponding collection of agents execute on a network of computer workstations. The agents operate within this framework asynchronously and, in general, autonomously. RAD permits the collection of agents to be dynamic, allowing agents to come and go. It accomplishes this by supporting a rudimentary learning capability, whereby

agents learn about each other when they receive messages. Rosette actors serve as communication aides, one representing each agent or database, to forward these messages (see Figure 2.2). They use a tree-space mechanism implemented through OSI and TCP/IP protocols. The actors also buffer messages when the agents they represent are busy. These actors exist as entities in an Extensible Services Switch or ESS (described in Section 2.4.2). Each ESS can be on a different host. Each ESS also contains a special actor, called aideServer, that maintains a partial directory of agent and database locations. The directory is updated when agents connect or disconnect from an ESS.

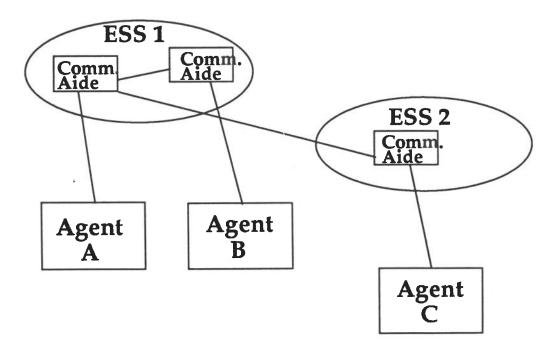


Figure 2.2: Each agent or database has an aide—a Rosette actor—that manages its communications

The design of the RAD framework is based on our observations of how a user interacts with a single, conventional knowledge-based system. A user can query and make assertions to the knowledge-based system, to which

the system responds with either an answer or an acknowledgement, as appropriate. As a side-effect of forward or backward inferencing, the system can independently inform or make a request of a user, who must answer all requests. Interactions can thus be either user-initiated or system-initiated.

Similarly, RAD agents communicate at a fundamental level by exchanging messages, specifically, queries and assertions. The agents can be either passive or active, i.e., they can either respond to questions and commands from another agent or initiate dialogs with another agent.

A problem is then solved by RAD in the following way. At any given time there will be a number of identically-structured computational agents, each having its own domain specific knowledge, and a number of users, each assisted by an interface. A problem, or a goal to be satisfied, is given to one of these agents by a user. This agent does one of the following two things:

- 1. Solves the problem or satisfies the goal itself, using local knowledge and internal reasoning mechanisms.
- 2. Decomposes the problem or goal into subproblems, distributes each one of these subproblems to an appropriate agent, and then integrates the solutions returned by these agents.

In either case, the agent reports the final solution to the user. Note that each of the other agents, upon receiving a subproblem, also does one of the above two things.

The features required for this type of distributed problem-solving include a communication channel among the agents, a communication protocol for exchanging goals and solutions on this channel, and a distributed reasonmaintenance system that enables globally-coherent solutions to be achieved. We have incorporated these features into the current version of RAD.

## 2.2 Architecture and Use

RAD extends this model to the multiple agent case by enabling systeminitiated interactions to be directed to another agent, as well as to a local user. For this purpose, a destination field is added to each query or assertion. A value for this field, i.e., the name of an agent who should receive the query

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<sup>&</sup>lt;sup>1</sup>At least for now.

or assertion, can then be the subject of a reasoning process in the originating agent. Since an assertion can now come from several sources, its justification must be modified to record a source. An agent must also record the agents to whom it sent an assertion, so that it can inform them in case its belief in that assertion ever changes.

Agents are implemented as a CLASS in RAD. This implies that all agents<sup>2</sup> have to be instances of the class AGENT. Agents have the following basic attributes:

Name:

<Agent Identifier>

Reliable:

<A boolean> Unreliable: <A boolean>

Said:

<List of assertions>

Was-told:

<List of assertions>

They store values of these attributes for themselves and each other. These attributes refer to a history of interactions with an agent and its trustworthiness. The agents exist in a fully distributed environment. Each agent can make queries of other agents and also serve as a knowledge base for other agents. That is, each agent works in a Client-Server mode. It is a Client when it requests other agents to solve a problem or satisfy a goal. It is a Server when it satisfies goals for other agents. An agent can be in one of the following three states:

- Busy
- Query
- Free

An Agent is in a BUSY state if it is performing internal inferencing operations. In this state, all requests and messages from other agents are buffered on its standard port. These requests are processed in order of arrival once the agent returns to the FREE state.

When an agent queries another agent, it can either go into a QUERY state until it receives a result or dispatch the request and carry on with other chores. The first choice refers to a serial way of solving problems and

<sup>&</sup>lt;sup>2</sup>Agents must have compatible class structures.

the second one refers to a parallel way of solving problems. In the current implementation, the agents operate serially. When an agent is waiting for a reply to a query, it will not permit other agents either to perform destructive operations on it or to make queries of it, as this may lead to a potential deadlock. Its communication aide buffers all messages until one with an awaited tag arrives.

Agents can also tell facts to other agents. This occurs asynchronously, and the receiving agent does not acknowledge the fact. This feature can be used to simulate *parallel* problem-solving that was mentioned above. To do so, the agents' knowledge must be expressed as forward rules, which automatically trigger on the assertion of facts in an agent. We have developed a declarative implementation of the *Contract Net* protocol [Smith and Davis 1981] based on this principle.

A user can also utilize the windowing system to view and edit the knowledge base of a remote agent from the user's terminal. The interface is described in detail in Chapter 9.

The following are the salient issues in distributed reasoning:

- Remote queries
- Remote assertions
- Remote justifications (to enable the implementation of a distributed JTMS [Mason and Johnson 1989])
- Negotiation

#### 2.2.1 Interagent Queries

An agent or user can make queries of other agents. This is one of the most basic operations in RAD. When an agent sends a query to a second agent, the second agent uses its reasoning mechanisms to find an answer. If it succeeds, then it returns its answer to the first agent, who records the answer, along with a justification that the source of the answer was the second agent. Meanwhile, the second agent remembers its answer, along with a justification that corresponds to the reasoning process it used and a note that it communicated this result to the first agent. The queries can be either simple

queries, i.e., queries at the user-interface level, or proof attempts of a literal of a rule at a remote agent.

More specifically, the goal to be proved is sent to the remote agent (the server) along with a Client Id. This registers a request at the Server's aide. Eventually, this request is forwarded to the Server, which then attempts to prove the given goal. If a successful proof is obtained, a datum corresponding to the proved literal is created. This datum has the justifications for the proof of the goal. This datum also is associated with a remote justification, which has a field to store the Client Id. Note that this datum participates as any other datum in the local JTMS. The operations described above are performed for each successful proof of the given goal. Finally all the proved literals are returned as a list to the client. Now if the server ever changes its beliefs about some literals or rules that result in a change in status of a datum, it checks to see if that datum has an associated remote justification. If it does, the server sends an update message to all the listed clients.

The above mechanism requires the server to remember every proof it performs for its clients and to remember their names. Sometimes, this is unnecessary and wasteful of storage and computation. Therefore, we have also provided the option of having a goal proven by a remote agent, where the remote agent does not remember the result. The former option is called **prove** and the latter option **query**.

When an interagent query with caching of results by the server occurs in a forward or backward rule, its syntax must be

```
prove(<Agent name> <goal>)
   or
prove-once(<Agent name> <goal>)
```

When an interagent query with no caching of results by the server occurs in a forward or backward rule, its syntax is

```
query(<Agent name> <goal>)
   or
query-once(<Agent name> <goal>)
```

An example of a command from a user to Agent1 to query tt Agent2 and Agent1's subsequent reply is (note that the syntax is different from the interface than from within a rule)

14

User: query Agent2 color(?x ?y)
Agent1: color(My-Ford Red)

color(My-VW Blue)

The messages that are exchanged by the communication aides encode the sender, the intended recipient, and a tag that uniquely identifies the message.

The Client now creates a datum of the type Client-Datum for each proof received from the server. This is an "image" of the server datum. Information about the server is stored in this datum. This datum is given a remote justification, which is a special kind of justification. Finally, this datum is stored in a local associative table at the client with the tag being the associative key. Note that this datum participates in the client's JTMS. Hence, the query could be a part of a rule. This datum is special, in that its status (IN/OUT) depends not only on local justifications, but also on proofs from remote agents. This is the justification for the above bindings. Finally, this datum is assigned the status IN.

#### 2.2.2 Remote Assertions

In RAD, we have provided agents with the capability to "tell" facts to other agents. This kind of interagent assertion, or remote assertion, is analogous to an assertion that a user might have made. Remote assertions are made using the Tell command, whose syntax is given below. Such a fact, when asserted on a remote agent, is justified by a remote justification. This remote justification has the pointer to the agent that made that particular assertion. Since any number of users or computational agents can communicate with an agent, assertions are justified by the source of the assertion, which may be a user or an agent.

The syntax for a remote assert is

tell(<hostname> <assertion>)

## 2.2.3 Remote Justifications and Distributed TMS

Remote data and justifications are created as described in the previous section. When the remote agent changes its belief about a datum, i.e., changes its belief status, and this datum happens to be of type Server-Datum, then

the corresponding Client, which is responsible for the creation of this datum, is notified. At the receipt of this message at the client, the "image" datum at the client is accessed. It is this datum that needs to be acted upon. If the transition is from an OUT to an IN, a new remote justification is created at the Client and is pushed onto the justifications of the client datum and the status of the datum is made IN. If the transition is from an IN to an OUT, the client datum is made OUT by an erase operation. Again, the justification for the subsequent ERASE datum is a newly created remote justification. The information about the agents causing the above changes is stored in the remote justifications. The Client also has an option of ignoring the messages from its servers, when the remote data it depends on change their status. This is set by a flag.

Note that the above remote justifications are necessary only for proofs whose justifications are needed. This may exclude simple user-level queries, whose proof tree (or justification information) is transient. An example where the justification is needed is remote proof attempts on one or more antecedents of a forward rule. These remote justifications will have to be entered in the justification of any assertion of the consequents of this forward rule.

#### 2.2.4 Negotiation

There is no requirement for two agents to agree completely. That is, each agent maintains a locally consistent set of beliefs using its own justification-based truth-maintenance system, but global consistency among the agents is not required. However, when two or more agents disagree about belief in a datum and when this disagreement is encountered during problem solving, then negotiation among the agents will ensue to resolve the disagreement. The negotiation is necessary to ensure that the global solutions to the problems posed to the agents are coherent.

The negotiation procedure involves an exchange of justifications among the agents. For example, if there is a disagreement about belief in datum  $D_1$ , then the agents exchange the justifications for their belief or disbelief in  $D_1$ . The agents then compare these justifications. If there is a disagreement about belief in any datum in the justifications, then the justifications for this disputed datum are exchanged, and the process recurses. It is expected, but not yet proven, that resolution of all disagreements among two agents in a

consistent world will be achieved by one agent informing another of some explicit knowledge that the other is missing. The negotiation process is an active area of research at MCC and will be described further in a forthcoming technical report.

## 2.3 Distributed Reasoning Example

Figures 2.3–2.5 show an example of the use and operation of our multiagent truth maintenance algorithm as two agents interact. Figure 2.3 shows the initial state of the knowledge bases for the two agents, Client, an investor, and Broker, a stockbroker. First, Client asks Broker to recommend a stock. Client knows what it can afford, but it does not have any knowledge about what stocks it would be wise to buy. It does know, however, that Broker knows about this problem. Broker does have some general knowledge about what stocks to recommend and about what stocks it can afford.

When Broker receives Client's query, it attempts to find an answer by applying its rules. It succeeds and binds tt ?X to XCorp. It also creates a server datum for this conclusion, containing the conclusion, the justification for the conclusion (rules 1 and 2), and a pointer to the agent to whom this conclusion was reported (Client). Broker recommends XCorp, which causes Client to believe that he should buy that stock, as shown in Figure 2.4. However, Broker then learns (not shown) that the basis for his recommendation, that XCorp is cash-rich, is no longer valid. He revises his beliefs and notifies Client that he has retracted his recommendation for XCorp. The final knowledge bases for the agents are shown in Figure 2.5.

It is interesting to note that these agents maintain different beliefs about whether or not they can afford to buy the stock of XCorp. Allowing this difference in belief—this global inconsistency—is useful in this case. It allows different viewpoints to be represented, it simplifies the representation of knowledge, in that the predicate afford really should have an additional argument indicating which agent can or cannot afford the stock, and it eliminates the interactions that would be needed to resolve the difference. Of course, the system would detect, and subsequently correct, the difference if the agents ever share or discuss this predicate. Details of the justifications that are constructed when two agents exchange beliefs are shown in Figure 2.6.

```
Client
    afford(XCorp Yes)
f3:
    Infer buy(?X) from
r3:
         query(Agent2 recommend(?X)) and
          afford(?X Yes)
                           recommend(?X)
                    Broker
     afford(XCorp No)
f1:
     cash-rich(XCorp)
f2:
     Infer recommend(?X) from takeover-bid(?X)
r1:
     Infer takeover-bid(?X) from cash-rich(?X)
```

Figure 2.3: Initial knowledge bases of two interacting multiagent-JTMS-based agents, before Client queries Broker for a recommendation

```
Client
    afford(XCorp Yes)
f3:
r3:
    Infer buy(?X) from
          query(Agent2 recommend(?X)) and
          afford(?X Yes)
    recommend(XCorp)
f4:
      Status - EXTERNAL
      Shared with - Agent2
      Justification - ()
f5: buy(XCorp)
      Status - IN
      Justification - (f3 f4 r3)
                        recommend(XCorp)
                   Broker
f1: afford(XCorp No)
f2: cash-rich(XCorp)
r1: Infer recommend(?X) from takeover-bid(?X)
r2: Infer takeover-bid(?X) from cash-rich(?X)
f3: recommend(XCorp)
      Status - INTERNAL
      Shared with - Agent1
      Justification - (f1 r1 r2)
```

Figure 2.4: Knowledge bases of the agents after Broker has replied to Client's query

```
Client
f3:
    afford(XCorp Yes)
    Infer buy(?X) from
r3:
          query(Agent2 recommend(?X)) and
          afford(?X Yes)
f4: recommend(XCorp)
      Status - OUT
      Shared with - Agent2
      Justification - ()
f5: buy(XCorp)
      Status - OUT
      Justification - (f3 f4 r3)
                        Relabel recommend(XCorp)
                   Broker
f1: afford(XCorp No)
f2: cash-rich(XCorp) → OUT
     Infer recommend(?X) from takeover-bid(?X)
r1:
r2: Infer takeover-bid(?X) from cash-rich(?X)
f3: recommend(XCorp)
      Status - OUT
      Shared with - Agent1
      Justification - (f1 r1 r2)
```

Figure 2.5: Final knowledge bases of the agents after Broker has notified Client that a fact must be relabeled

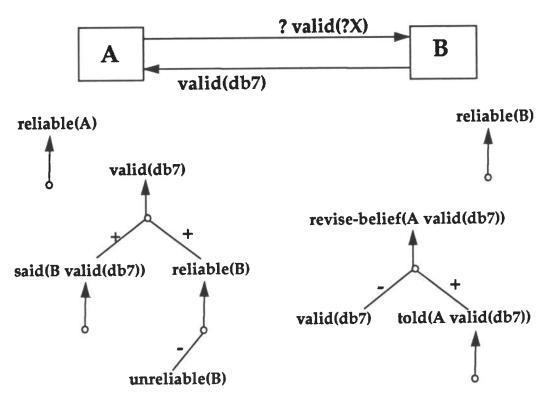


Figure 2.6: Justification networks that are constructed by two agents that exchange beliefs (see Chapter 3 for notation)

## 2.4 Implementation Issues

#### 2.4.1 Protocol for Message Passing

The communication aides utilize the Rosette tree-space mechanism to pass messages. This is shown in Figure 2.7. The agents establish TCP/IP connections from the machine on which they are executing to a machine running an Extensible Services Switch or ESS (please see Section 2.4.2 for details). Establishment of a connection causes the creation of a communication aide and a tree-space interface for each agent.

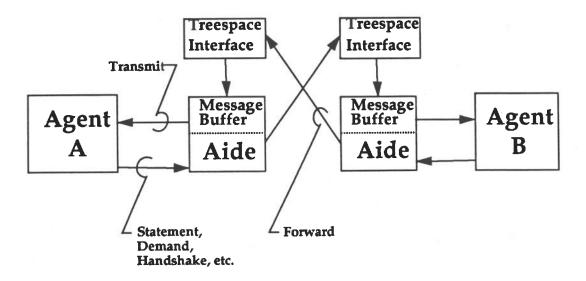


Figure 2.7: Each communication aide has an associated tree-space interface that passes messages for agents or databases

Once a connection is established, agents and databases can exchange six basic types of messages: demands, responses, database queries, statements, errors, and handshakes. Each of these message types encodes information about the sender, the communication aide, the recipient, the identifying tag, and the contents. The message types and their specific behavior are described next.

Demand: a message for which the sender expects a response to be returned. Queries between agents are *Demands*, but so are assertions and com-

mands from a user, since they require an acknowledgement. The semantics of a demand is that the agent will not assume that the remote agent has performed the operation until a response has been received. A demand message has the following syntax:

(outDemand <my-aide> <destination> <new-tag> <message>)

The local communication aide forwards this message through the tree space to its destination and then waits for a response with a tag that matches <new-tag>. While it is waiting, it buffers all other messages for its agent. This means that agents will remain idle until they receive a response to their demand. Interfaces, however, can continue to send out other messages.

Database Demand: a message to be sent to a database. The message contents are an SQL statement to be processed by the database. It has the following syntax:

(Eval (db-query DBAgent <SQL-statement>[])[<destination>])

Since databases are passive entities, it convenient to let the response of a database demand arrive as a result of computing the given db-query on the destination.

**Response:** a message in response to a *Demand* message. It uses the tag from the demand to which it is responding. It has the following syntax:

(outResponse <my-aide> <destination> <old-tag> <message>)

Statement: a message from one agent to another, where a reply is not expected. This is an asynchronous input to an agent. An example of a *Statement* is a print message to the interface. A statement has the following syntax:

(outStatement <my-aide> <destination> <new-tag> <message>)

Error: a message issued as a response to a *Demand* message, when the *Demand* message caused an error in the agent. It has the following syntax:

#### (outError <my-aide> <destination> <old-tag> <message>)

Handshake: a message that an agent sends to its communication aide, informing it that the agent is FREE and is ready for the next message. It has the following syntax:

#### (handshake <my-aide>)

In order for the agents and interfaces to be robust in the presence of unexpected messages, expecially when they are waiting for a response to one of their demands, we have implemented message handling by the communication aides according to the state diagram shown in Figure 2.8. The state machine specifies the operational semantics of messages that are sent and received by agents and interfaces.

## 2.4.2 Extensible Services Switch (ESS)

Each agent must be connected to an ESS. The ESS then provides communication facilities and a simple directory mechanism for maintaining the locations of agents, databases, and other ESS's.

Whenever an agent is started, it registers itself with an ESS. The syntax for registration is

#### (makeAide aideServer <Aide Name> <Agent Name>)

Whenever an agent halts or ceases to exist, it unregisters from the ESS by sending the message

## (killAide aideServer <Agent Name>)

At any time, an agent can unregister from one ESS and then register with another. The other messages supported by the ESS for the purpose of agent interactions are

- · db-connect, for connecting a database to an ESS,
- db-disconnect, for disconnecting a database from an ESS,
- essOf, for determining whether a given agent is registered and finding its ESS, and

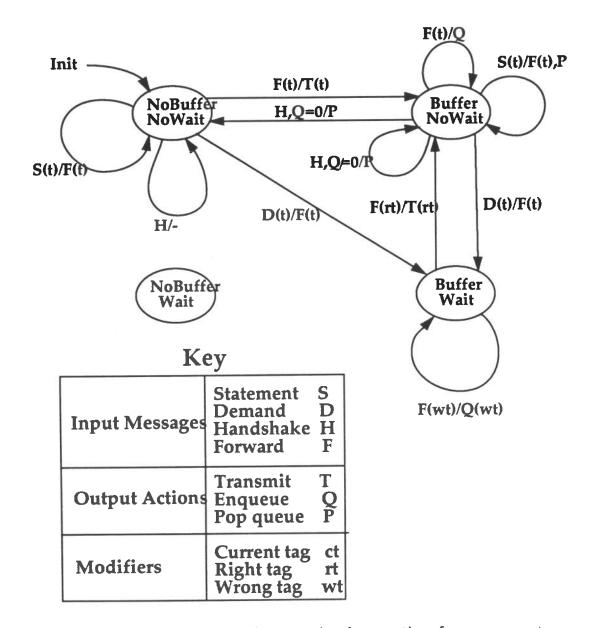


Figure 2.8: A state machine for the operational semantics of messages sent and received by computational agents and interfaces

• who, for producing a list of all agents and databases, and their ESS's known to an ESS.

The remainder of this document primarily describes the behavior, functionality, and structure of a single agent.

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# Chapter 3

# Truth Maintenance

Each datum (or element of a RAD knowledge base) represents a potential belief. The status of this belief, which is subject to change, is reflected in the *support-status* of the datum, the value of which may be IN, indicating that it is currently believed, or OUT, indicating current disbelief. This value is assigned by a justification-based truth-maintenance system (JTMS) in accordance with a list of *justifications* that have been attached to the datum.

Each justification consists of a pair of lists of data, the IN-list and the OUT-list of the justification. A justification is said to be valid and is considered to represent reason for belief in its associated datum if each element of its IN-list is IN and each element of its OUT-list is OUT. The justified datum is said to depend monotonically on each member of the justification's IN-list and nonmonotonically on each member of the OUT-list.

Also associated with each datum is a list of other data called its *sup*porters. The supporters of a datum are considered to be responsible for its current support-status.

It is the function of the JTMS to assign support-statuses and supporters to data in a manner that is consistent with their justifications, and to adjust these assignments continually as required by the addition of new justifications and the retraction of old ones. More precisely, the state of the knowledge base, as constructed by the JTMS, must satisfy two requirements: stability and well-foundedness. A stable state is one that satisfies the following conditions:

1. A datum is IN if it has at least one valid justification. In this case its list of supporters is the result of appending the IN-list and OUT-list

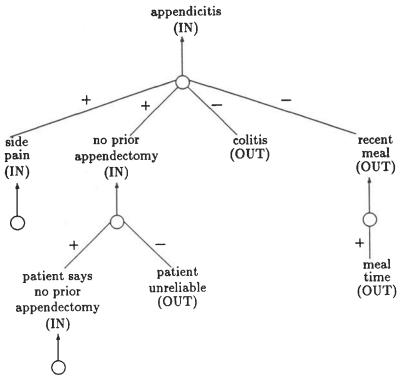


Figure 3.1: A stable well-founded state

of one of its valid justifications. This justification is identified as the supporting justification.

2. A datum is OUT if it has no valid justification. Its supporters then include one representative of each of its invalid justifications: either an OUT member of the IN-list or an IN member of the OUT-list.

The requirement of well-foundedness is that no set of beliefs be mutually dependent, i.e., there may be no sequence of data  $d_0, \ldots, d_n$ , all of which are IN, such that  $d_0 = d_n$  and for  $i = 1, \ldots, n, d_{i-1}$  is a supporter of  $d_i$ .

An example of an admissible state is shown in Figure 3.1. In this graph and those that follow, each circle corresponds to a justification, with an arrow pointing to the justified datum, positive arcs connected to the elements of the IN-list, and negative arcs to elements of the OUT-list. Thus, the datum representing a diagnosis of appendicitis has a valid justification with a two-element IN-list and a two-element OUT-list. The belief that the patient has a

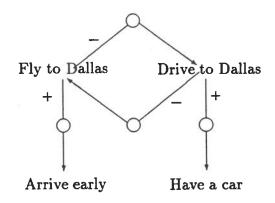


Figure 3.2: Alternative assumptions

side pain is supported by a justification with an empty IN-list and an empty OUT-list and is said to be a premise. The datum representing the unreliability of the patient has an empty list of justifications and is therefore OUT. If this datum were to acquire a new valid justification, then its support-status as well as those of the data that depend on it (directly or indirectly) must be reevaluated, ultimately forcing the diagnosis OUT. This phenomenon, the development of a new belief resulting in the abandonment of an old one, characterizes nonmonotonic reasoning.

In the presence of nonmonotonic dependencies, the status-assignment problem may not have a unique solution. In the situation shown in Figure 3.2, the JTMS may succeed either by making Fly-to-Dallas (and hence Arrive-early) IN and Drive-to-Dallas (hence Have-a-car) OUT, or by giving the opposite assignments. This choice between alternative hypothetical assumptions can only be made arbitrarily, and may have to be revised later as new justifications are produced (e.g., if Drive-to-Dallas acquires a new valid justification while Fly-to-Dallas is IN).

Circularities involving nonmonotonic dependencies may also impose unsatisfiable constraints on the JTMS, a situation that may be difficult to detect. Two simple examples of this are shown in Figures 3.3 and 3.4. Note that the network of Figure 3.4 does have a stable state (in which all data are IN), but this state is ill-founded, and therefore inadmissible.

Figure 3.5 illustrates a situation in which a datum Q is given a new justification and, in order to restore stability, the JTMS must examine not only Q and the data that depend on Q, but also those on which Q depends. In this example, if only the support-statuses of P and Q were reassigned, the system would be forced to report an unsatisfiable circularity. In order

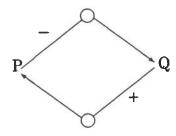


Figure 3.3: Unsatisfiable dependencies

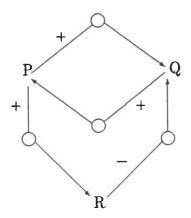


Figure 3.4: No well-founded stable state

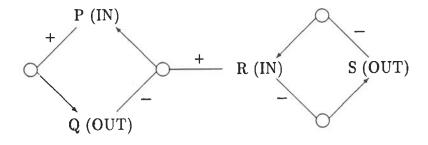


Figure 3.5: A solvable odd loop

to succeed, it must also consider R (on which Q now depends) and thus S (which depends on R). The only solution is to put S IN and P, Q, and R OUT.

As described in [Russinoff 1985], the RAD JTMS is complete in the sense that given any database with any set of justifications, it will achieve a stable well-founded state if such a state exists, and otherwise will recognize and report failure. This represents an improvement over the original TMS [Doyle 1979], as well as other published procedures for truth maintenance [Charniak et al. 1980, Goodwin 1984]. These systems all fail (perhaps even fail to terminate) in the presence of certain circular dependencies that have been characterized as odd loops. An odd loop is a cycle of arcs with an odd number of minus signs, as in Figures 3.3, 3.4, and 3.5. A dependency network containing such a loop may or may not be satisfiable. (We have already observed that the network in Figure 3.5 does admit a solution.) While the presence of odd loops complicates the truth maintenance task and is generally considered undesirable, they are sometimes unavoidable in practice, particularly in dependency networks that are based on input from several users.

In the following chapters, we shall see how data dependency networks are created in *RAD* by the user, by the frame system, and by both forward and backward inference.

# Chapter 4

# Frames and Assertions

The data on which the JTMS operates represent statements about objects. Before discussing the structure of these data, we shall describe the objects that they concern. These objects, called *frames*, are the subject of this chapter.

## 4.1 Classes, Subclasses, and Instances

In the initial state of the system, there exist several frames. One of these, named CLASS, plays a special role as discussed below. The others are AGENT, DATABASE, RELATION, LIST, CONS, NULL, SYMBOL, NUMBER, FIXNUM, FLOAT, BIT-VECTOR, VECTOR, and STRING. By means of the commands described in Chapter 9, the user may enlarge this set by creating new frames.

There are two primitive relations defined on frames: Instance and Subclass. If a pair (x, y) is an element of the instance relation, we say that x is an instance of y, or that y is the type of x. For a pair (x, y) in the subclass relation, we say that x is a subclass of y or that y is a superclass of x. Figure 4.1 depicts these relations as they are defined in the initial state. Broken lines are drawn from instances to types; solid lines are drawn from subclasses to superclasses. Note that CLASS is the type of every system-defined frame.

Two other important relations are defined in terms of these primitives:  $Subclass^*$  and  $Instance^*$ . The relation  $Subclass^*$  is defined as the reflexive transitive closure of the relation Subclass. Thus, x is a  $subclass^*$  of y (equiv-

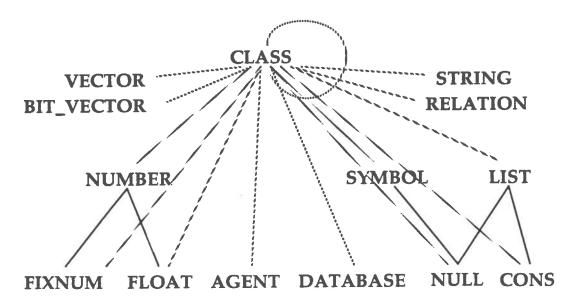


Figure 4.1: Built-in classes. Solid lines denote subclass relationships and dashed lines denote instance relationships.

alently, y is a superclass\* of x) if either x is identical to y, or x is a subclass of a subclass\* of y. The relation Instance\* is defined as follows: x is an instance\* of y iff x is an instance of a subclass\* of y. In this case we may also say that x is a y. In particular, a class is by definition a frame that is an instance\* of CLASS. Note that every system-defined frame is a class, including CLASS itself.

As new frames are created by the user, he may also extend these relations by assigning types to instances and creating subclass relationships (between user-defined classes only). This must be done in such a way, however, that at each stage of the development, the following properties are preserved:

- 1. The instance relation is a function, i.e., for each frame x there exists a unique frame y such that x is an *Instance* of y.
- 2. CLASS is the only frame that is an *Instance* of itself. Thus, whenever a new frame is created, some preexisting frame must be specified as its type.

- 3. The relation Subclass\* is a partial order. That is, if x is a Subclass\* of y and y is a Subclass\* of x, then x and y are identical.
- 4. If x is an Instance of y, then y must be an Instance\* of CLASS.
- 5. If x is a Subclass of y, then x and y must each be an Instance\* of CLASS.

Thus, according to the last two of these properties, only a class may have a subclass, superclass, subclass\*, superclass\*, instance, or instance\*. There is a further restriction on the classes that may be instantiated by the user: a user-defined frame may be an instance of CLASS, RELATION, or of any user-defined class, but it may not be an instance of LIST, CONS, NULL, SYMBOL, NUMBER, FIXNUM, BIT-VECTOR, VECTOR, STRING, or FLOAT. Instead, whenever the system encounters a Common Lisp object whose datatype is the name of one of these system-defined classes, the object automatically becomes an instance of the named class. For example, if the number 3 is read, it becomes an instance of FIXNUM and thus a member of NUMBER (in other words, a number). When the symbol NIL is encountered, it is recognized as the unique instance of the class NULL, and hence both a list and a symbol.

An example of a user-defined system of frames is illustrated in Figure 4.2. This example involves eight new classes, all of which are subclasses of the class PERSON and instances of the class CLASS. For clarity, classes are denoted in bold-face and other user-defined frames in typewriter font. HILARY, for example, as an instance of TA, is not a class, but is a TA, a graduate, a staff, a student, an employee, and a person.

## 4.2 Metaclasses

Of course, the existence of classes as frames provides the advantage of being able to reason about classes at the same level at which one reasons about the objects of which they are comprised. It is often desirable to be able to reason about classes of classes as well. The only class we have seen so far that has classes as instances is CLASS itself. Any class that has a class as an instance must be a subclass of CLASS, in which case all of its members are

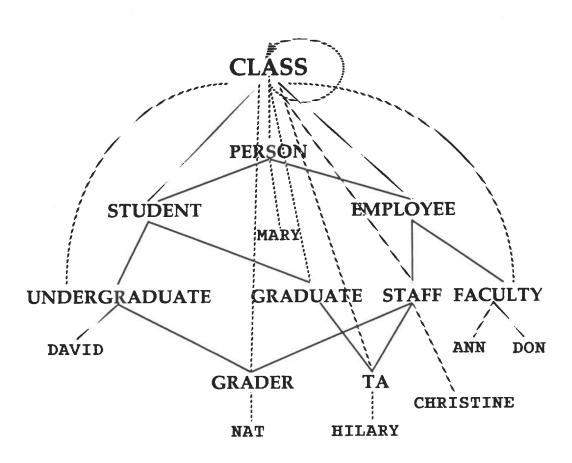


Figure 4.2: User-defined classes

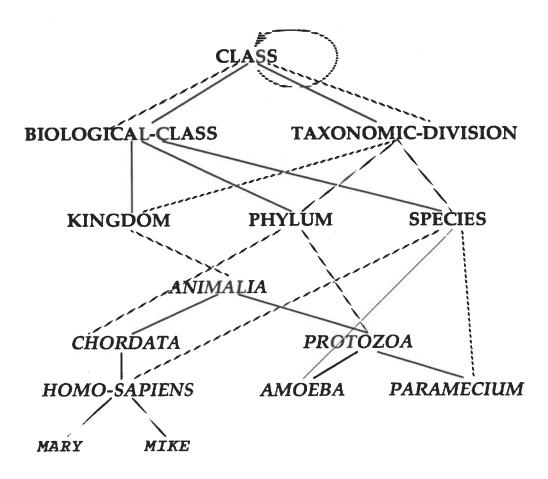


Figure 4.3: User-defined metaclasses

classes. Such a class is called a *metaclass*. While **CLASS** is the only system-defined metaclass (and the only one in the example of Figure 4.2), additional metaclasses may be created simply by asserting a subclass relationship from any user-defined classes to **CLASS**.

The example of Figure 4.3 includes several user-defined metaclasses (denoted in large bold print). This example is based on six classes of animals: ANIMALIA (the class of all animals) and five of its subclasses, which are related as indicated by the solid arrows connecting them. These classes could be constructed simply as instances of CLASS. But in order to represent and utilize the knowledge that these classes share more

than mere classhood, we first define a metaclass called BIOLOGICAL-CLASS, intended to include the animal classes among its instances. In fact, the animal classes are partitioned into smaller metaclasses by defining them as instances of KINGDOM, PHYLUM, and SPECIES, which are subclasses of BIOLOGICAL-CLASS. Note that KINGDOM, PHYLUM, and SPECIES are themselves not simply instances of CLASS, but are classes by virtue of being instances of the metaclass TAXONOMIC-DIVISION. Moreover, they are metaclasses by virtue of their links to the metaclass BIOLOGICAL-CLASS.

## 4.3 Variables and Types

Along with the objects that appear in RAD data, there are also occurrences of variables. These are denoted as symbols with initial character "?". Variables, as usual, represent unspecified objects. The process of unification, which is central to the mechanisms of forward and backward chaining, is built on the basic operation of binding (i.e., assigning values to) variables. An unbound variable ?X may be bound either to an object or to another unbound variable ?Y. In the latter case, if ?Y is subsequently bound, its binding also becomes the binding of ?X.

Since objects in this system are classified by their types, it is natural and useful to classify variables as being of a certain type [Aït-kaci, et al. 1985]. This is done by appending the name of a class to the variable name, using ":" as a separator, as in ?X:STUDENT. The consequence of assigning a type to a variable is that any binding of the variable is required to be an instance of the variable's type.

In the presence of typed variables, the standard unification algorithm must be altered in several ways. First, before a variable is bound to an object, it must be verified that the object is an instance of the variable's type. Second, before a variable is bound to another variable, it must be verified that the types of the two variables are compatible, so that it will be possible later to bind them to the same object. Thus, the two types must have a nontrivial common subtype. Finally, when a variable ?X is bound to a variable ?Y, the type of ?Y must be replaced in order to ensure that any later binding of ?Y is consistent with the type of ?X. The new type of ?Y should be the most general common subtype, or greatest lower bound, of the

type of ?X and the old type of ?Y. For example (see Figure 4.2), a variable ?X:UNDERGRADUATE could be bound to a variable ?Y:STAFF, with type of ?Y replaced by GRADER. ?Y could then be bound to NAT, but no longer to HILARY, which would violate the type restriction on ?X.

Thus, the modified unification algorithm depends on the computability of the greatest lower bound (g.l.b.) of two variable types. Unfortunately, the partially ordered set of classes does not form a *lattice*, i.e., the g.l.b. of two classes may not exist. The classes **STUDENT** and **EMPLOYEE** of Figure 4.2, for example, have two common subclasses, **GRADER** and **TA**. Since neither of these is a superclass of the other, neither can be said to be the g.l.b. of **STUDENT** and **EMPLOYEE**.

This problem is solved by generalizing the notion of type. A type is now defined to be a set of classes, none of which is a subclass of another. A type  $t_1$  is a subtype of a type  $t_2$  if each of the classes of  $t_1$  is a subclass of at least one of the classes of  $t_2$ . An object is said to belong to a type if it is an instance of at least one of its classes. It follows that an object belongs to a type if it belongs to any of its subtypes.

Under this ordering, the set of all types forms a lattice. The type of a variable may be any instance of this lattice. The identification of each class c with the type  $\{c\}$  induces an embedding of the partially ordered set of classes into the lattice of types. In this context, the g.l.b. of two classes may always be computed. Thus, the g.l.b. of **STUDENT** and **EMPLOYEE** is the type  $\{GRADER, TA\}$ .

The empty set of classes, denoted BOTTOM, is a subtype of every type. This type contains no objects and is not allowed as the type of a variable. If the g.l.b. of the types of two variables (e.g., ?X:STAFF and ?Y:FACULTY) is BOTTOM, then these variables cannot be unified; their types contain no common objects.

The set of all maximal classes is also a type, denoted **T**. It is a supertype of every type, and every object belongs to it. If no type is specified for a variable, then the variable's type is taken to be **T** as a default. There is no restriction on the binding of such a variable.

The cost of the expressive power of typed variables is the resulting complication of the unification algorithm. In order to minimize this cost, the g.l.b. operation must be a fast computation. This is accomplished by means of an encoding scheme that associates with each class c a bit-string  $\mathcal{B}(c)$ , in such a way that  $c_1$  is a subclass of  $c_2$  if and only if  $\mathcal{B}(c_1)$  is (bit-wise) less

than or equal to  $\mathcal{B}(c_2)$ . For a type  $t = \{c_1, \ldots, c_k\}$ ,  $\mathcal{B}(t)$  is constructed as the logical-or of the  $\mathcal{B}(c_i)$ . The g.l.b. operation then reduces to logical-and. The details of this scheme are described in [Aït-kaci, et al. 1985].

## 4.4 Assertions

There are two main classifications of data: rules and assertions. Rules, which are used by the system to derive assertions from other assertions, are classified as forward rules and backward rules. These are the subjects of Chapters 6 and 7. Assertions, as described in this section, represent simple statements. Specifically, an assertion is a proposition that resides in the knowledge base, where a proposition is a list whose members are a relation symbol followed by arguments.

An assertion is further classified as *general* or *particular* according to whether or not it contains any variables. The variables in a general assertion are understood to be universally quantified. Thus, the assertion

User: owes(?U:undergraduate ?F:faculty 15)

represents the statement "Each undergraduate owes each faculty member 15 dollars."

### 4.4.1 Instance Slot Values

A frame may assume values for a given relation if the frame is an instance of the class for which the relation is defined (see Chapter 5). In this case, an instance slot corresponding to the relation is attached to the frame. One or more values may be stored in this instance slot. Thus, if a relation Description has argument-types PERSON and HUMAN-ATTRIBUTE, then the frame NAT may acquire Description values. An assertion such as

User: description(nat tall)

Zeus: Instance Slot Value DESCRIPTION-1 accepted.

results in the creation of a datum called an *Instance Slot Value*, which is stored in NAT's *Description* slot. This datum is justified as a premise (i.e., with empty IN-list and OUT-list). Thus, an instance slot value is just an assertion pertaining to a relation for an instance of a class.

When a relation is initially defined, it is either specified as Single-valued or is multiple-valued by default. A frame may assume any number of coexisting values for a multiple-valued relation. If Description, for example, is multiple-valued, then asserting a new Description for NAT via

User: description(nat thin)

has no effect on the old value. Thus, a query for NAT's Description produces both values:

User: ?? description(nat ?X)
Zeus: DESCRIPTION(NAT TALL)
DESCRIPTION(NAT THIN)

For a single-valued relation, on the other hand, a frame may have only one effective value at any time. Suppose the relation *Nationality* is defined for class **PERSON** and declared to be single-valued, and that a *Nationality* is asserted for NAT:

User: nationality(Nat French)

Zeus: Instance Slot Value NATIONALITY-1 accepted.

If another value is later asserted for the *Nationality* of NAT, then the old value is overridden by the new one:

User: nationality(Nat Swiss)

Zeus: Instance Slot Value NATIONALITY-2 accepted.

User: ?? nationality(Nat ?X)
Zeus: NATIONALITY(NAT SWISS)

Actually, this restriction to a single value is enforced by the JTMS: whenever a value is asserted for a single-valued relation, it is added to the OUT-list of each justification of any preexisting conflicting value. Thus, the old value remains in the database, but is ignored in answering the query because it is now OUT. This method not only provides for the construction of explanations, such as

User: why NATIONALITY-1

Zeus: Instance Slot Value NATIONALITY-1

#### JUSTIFICATION:

in-list NIL
out-list (NATIONALITY-2)

#### is invalidated by

Instance Slot Value NATIONALITY-2

but also allows an old value to be reinstated if the overriding value is removed:

User: erase nationality-2

Zeus: OK.

User: ?? nationality(nat ?X)
Zeus: NATIONALITY(NAT FRENCH)

In fact, RAD always ensures that the value that is IN is the one with the most recently created justification that is currently valid. Note that if several conflicting values for a slot are asserted in succession, then a reassertion of the original value will result in a complicated dependency network, including odd loops. This implementation, therefore, requires a complete JTMS as discussed in Chapter 3.

Instance slot values do not have to be binary. For example, consider representing "Nat owes David \$5." This would be written in RAD by using the ternary relation Owes as follows:

User: assert-instance (relation owes)

User: argument-types(owes (person person number))

User: single-valued(owes)
User: owes(nat david 5)

Further, the default that "Nat owes everybody \$10" could be written

User: owes(nat ?X:person 10)

### 4.4.2 Class-Slot Values

Values for a relation may be attached to subclasses of its defining class as well as to its instances. Each of these subclasses contains a class slot, the

values in which may be inherited by any instance of the class to which it belongs. For example, the command

User: description(?X:person aerobic)

Zeus: Class Slot Value DESCRIPTION-1 accepted.

adds a value to the *Description* class slot of **PERSON**: This datum represents the belief that every instance of **PERSON** is **AEROBIC**. Similarly, any subclass of **PERSON** may acquire class values for this relation:

User: assert-instance (class texan)
User: assert-subclass (person texan)
User: ?? description(?X:texan ?Y)
Zeus: DESCRIPTION(?X:TEXAN AEROBIC)

The *Description* values associated with an instance of **TEXAN** are then the instance values specifically assigned to it, along with the class values assigned to the superclasses of its type. Hence,

User: assert-instance (texan connie)
User: description(connie pretty)
User: ?? description(connie ?X)
Zeus: DESCRIPTION(CONNIE PRETTY)
DESCRIPTION(CONNIE AEROBIC)

For single-valued relations, a more complicated mode of inheritance is used. A value for a single-valued class slot may be inherited by an instance of the class only if no value has been assigned to that instance's instance slot. Class slot values for these relations are therefore called *default values*.

Default values assigned to a given class override each other in the same manner as instance values for a given frame, so that only one default value assigned to a class may be IN at any time. If a value for a relation is sought for a given frame, the frame's instance slot is first examined for an IN value. If there is none, then each of the superclasses of the frame is examined (in depth first order) until an IN default value is found.

Suppose that for the single-valued attribute Nationality, the default values AMERICAN, CHINESE, and INDIAN are asserted for the classes PERSON, STUDENT, and GRADUATE, respectively. Suppose further that NAT's Nationality is asserted to be FRENCH and that DON's is GERMAN. Then a query

for the *Nationality* values for all instances of **PERSON** would produce the following:

User: ?? nationality(?X:person ?Y)

Zeus: NATIONALITY (DON GERMAN)

NATIONALITY (ANN AMERICAN)
NATIONALITY (HILARY INDIAN)
NATIONALITY (NAT FRENCH)
NATIONALITY (DAVID CHINESE)
NATIONALITY (MARY AMERICAN)

If a new default value were now asserted for **STUDENT**, only **DAVID**'s *Nationality* would change:

User: nationality(?X:student carolinian)

User: ?? nationality(?X:student ?Y)

Zeus: NATIONALITY(HILARY INDIAN)

NATIONALITY (NAT FRENCH)

NATIONALITY (DAVID CAROLINIAN)

If the default value for **GRADUATE** were retracted, then *HILARY* would inherit from **STUDENT**:

User: erase nationality(?X:graduate indian)

User: ?? nationality(?X:student ?Y)
Zeus: NATIONALITY(HILARY CAROLINIAN)

NATIONALITY (NAT FRENCH)

NATIONALITY (DAVID CAROLINIAN)

# Chapter 5

# Relations in RAD

## 5.1 Relations as Objects

In RAD, an object is a cluster of knowledge, i.e., a thing to which attribute values are attached and about which propositions are formed and inferences are drawn. An object is also a thing that is subject to a classification according to which it inherits properties. Of all the structures that are manipulated by an inference system, which ones are to be handled explicitly as objects in this sense?

In an object-oriented or frame-based system, any structure about which one cares to reason should have the status of an object. In Proteus, the only built-in objects are a few system-defined classes. Predicates are not objects in the above sense, and neither are rules, assertions, or slots. These structures are allowed to occur as arguments of predicates, but they may not have attribute values and they cannot inherit properties. This has proved to be a deficiency—the machinery that has been developed for classifying and reasoning about objects is not available in dealing with these less fortunate structures, even though it might be quite useful. Raising their status to make them objects seems to be desirable in terms of the power of our system, the simplicity of its implementation, and the clarity of the user's model. Thus, the following are full-fledged objects:

Classes Every object is an instance of at least one class. If classes are objects themselves, then they must be instances of other classes. These classes of classes are called *metaclasses*, and they can be manipulated in the

same ways as ordinary classes.

Relations Conceptually, a relation of arity n is a set of n-tuples of objects. In Proteus, these are partitioned into "predicates" and "attributes." A predicate may have arbitrary arity and there are no restrictions on the form of its arguments. An attribute is always binary, and has a well-defined domain. That is, its first argument is constrained to belong to some specified class for which the attribute is defined. Thus, an attribute allows values to be attached to objects—it may be thought of as a mapping from some domain of objects into some set of values. If attributes (more generally, relations) themselves have the status of objects, then properties such as reflexivity, symmetry, etc. may be explicitly associated with them. It would also allow a classification of relations according to mode of inheritance and other properties.

The structure of a frame system is defined by the primitive relations among its objects. Each of these relations may be viewed as a built-in attribute, defined for some class of objects. Some of the most obvious of these are

- 1. Type, defined for the universal class of all objects;
- 2. Instance and Subclass, defined for the class of all classes;
- 3. Argument-types and Single-valued, defined for the class of relations;

Other built-in attributes represent the links between attributes and slots, slots and objects, rules and rule instances, objects and their parts, etc. One advantage of representing these primitives as attributes is that it allows the structure of the frame system to be defined within the frame system itself. Another advantage is that if instance and subclass links (which comprise the paths through which properties are traditionally inherited) are special cases of slot values, then it begins to make sense to think about supporting inheritance through arbitrary relations. Relations in RAD have three significant characteristics:

• In RAD there is a single class of object—relation—that is of arbitrary arity. A given relation can either be Specificity-ordered, in which case it does support inheritance, or not.

- In RAD relations are objects. This allows inheritance of information to relations.
- In RAD, there is a class **RELATION**. Every relation is an instance of that class.

## 5.2 Relations on Relations

Most of the following relations on relations are lazy: e.g., a relation can be asserted to be Axiomatic, and if later that assertion is killed, then the relation will no longer be Axiomatic. A few are not lazy and once asserted are permanent, such as Specificity-ordered, Single-valued, and (to some extent) Argument-types. The restrictions that apply are explained in detail with each of these three relations. Note that it is only difficult, but not impossible, to implement these three relations lazily.

#### Axiomatic reln

Primitive Axiomatic

Most relations have instances that are objects in the JTMS, i.e., an assertion about such a relation would have a support status and justifications. However, the instances of some relations are not JTMS objects, because it would be cumbersome and inefficient to tag everything that depended on them. Such relations are said to be axiomatic. For example, Argument-types (defined below) is axiomatic; if it were not, a premise assertion of the backward rule

Infer foo(?X) from ( bar(?X) baz(?X) )

would have to be justified with the facts that Foo, Bar, and Baz all take the argument types that they in fact do.

Note that although axiomatic relations can be single-valued, it is an error to assert an "extra" value for one. (This error handling is not yet implemented.)

Primitive reln

System Axiomatic

Some relations are *Primitive*, in that users can only make ground unit assertions with these relations. This restriction is imposed because some relations (e.g., *Subclass*) are handled specially by *RAD* machinery that assumes only ground unit assertions. If the user attempts to assert nonground

or nonunit classes about reln, RAD will warn the user of his mistake and refuse to accept the assertion.

System reln

System Axiomatic

Some relations are completely defined by the system. For example, Subclass\* is defined as the transitive closure of Subclass, and it is a mistake for the RAD user to attempt to assert clauses about Subclass\* directly. Such attempts will be rejected with an error message, because Subclass\* is a System relation.

A relation cannot be both primitive and a system relation, although this is not currently enforced.

Action reln

System Axiomatic

The group of relations that perform actions are designated by the unary relation *Action*. Action relations are typically used as consequents in forward rules.

Argument-types reln list-of-types

Primitive Axiomatic

Each relation has a fixed arity, and restrictions on the types of its arguments. List-of-types is a list of types of logical variables. A type can be either a class, a conjunction of classes (written as a list of the classes), or the unrestricted type (written as T).

The arity of reln is the length of list-of-types, and only tuples of arguments that unify with the types in list-of-types are permitted as tuples of arguments to the relation. Thus Argument-types generalizes the Proteus domain and range constructions to all the arguments of a relation.

Operationally, Argument-types has the following effects:

 An assertion that is not consistent with the Argument-types assertion will be rejected with an error message. For example if

and a user asserts

power-consumption(fan-25 100 joules)

RAD will reject the assertion because the user has the wrong units.

• Only proofs that are consistent with Argument-types assertions will be found. Thus, no proofs will be found of power-consumption(zebra?x?y).

Note that because Argument-types is multidirectional, it is also good for queries of what relations are defined on an argument of some type. For example, one can determine the relations defined on fan-25 by querying argument-types(?x (fan . ?ignore)).

If argument-types is not specified for a relation, but the arity of the relation is deduced from other knowledge, the default argument-types is a list of unrestricted types (T) of the right arity. The argument-types of a relation can be changed, but one cannot change the arity of a relation after it has been set.

#### Persistent reln

Primitive Axiomatic Single-valued

Persistent relations survive a clear operation on the knowledge base. There are two restrictions:

- 1. The only persistent classes are the built-in ones CLASS, AGENT, RELATION, FIXNUM, etc. Hence, a *Persistent* relation cannot refer to any other class.
- 2. A Persistent relation cannot be defined in terms of relations that are not Persistent. If F00 is defined by "Infer foo(?x) from bar(?x)" and F00 is Persistent, than BAR must also be.

Note that there is currently no error-checking for either of these.

All library routines of built-in predicates are persistent, so they do not have to be automatically reloaded after clearing the knowledge base. Note that since classes cannot be made persistent, this limits the use of typed variables in defining persistent relations.

### Single-valued reln

Primitive Axiomatic

A Single-valued relation is one that has only a single set of second through nth ground arguments for a ground first argument. As in Proteus, the latest assertion takes precedence over earlier ones. Single-valuedness is implemented with the JTMS.

A relation cannot be asserted to be Single-valued after clauses of the relation have been asserted. In this way Single-valued is not lazy. Similarly,

the Single-valued property of a relation cannot be retracted after the relation has clauses.

#### Specialization special-reln general-reln

Primitive

Special-reln is a specialization of the relation General-reln. The two relations must have the same arity, and the argument types of Special-reln must be nonstrict subclasses of the argument types of General-reln. The built-in relations that are specializations of each other are specialization(subclass subclass\*) and specialization(instance instance\*).

A proof of Special-reln for some arguments will suffice to prove it for General-reln for the same arguments.

#### Specificity-ordered reln

Primitive Axiomatic

Relations that support inheritance are *Specificity-ordered*. These relations must have an arity > 0 and must have a domain that is an actual class, not just a type. Unlike Proteus, inheritance in RAD is not restricted to binary relations.

A relation cannot be asserted to be *Specificity-ordered* after clauses of the relation have been asserted. In this way, *Specificity-ordered* is not lazy. Similarly, the *Specificity-ordered* property of a relation cannot be retracted after the relation has clauses.

## Triggerable reln

This probably should be a relation, but it currently is not.

## Unidirectional reln in-arg-indices

Primitive Axiomatic

Most relations in RAD are, by default, multidirectional. That is, any argument can potentially be used either for input or for output. However, some relations are either inherently or conceptually Unidirectional, meaning that the bindings for some arguments cannot (or should not, due to considerations of efficiency or conceptual clarity) be inferred given bindings for some others.

Unidirectional means that any query on reln must have ground bindings for all the arguments given in in-arg-indices. From an implementation

<sup>&</sup>lt;sup>1</sup>We could, of course, alter the semantics (or provide another relation) to make a weaker statement about the relation involved, namely, that if any of the arguments specified by *in-arg-indices* is nonground, then all arguments not in *in-arg-indices* are nonground as well. This intuitively does not seem as useful, although in such cases there would still be room for some optimization.

standpoint, the advantage of having some relations be unidirectional is that it could allow us to index instances of the relation on an argument or arguments. We could thus generate more efficient WAM/Lisp code for such relations.

Unidirectional has been defined here as a primitive relation, although most users would not actually want to concern themselves with the generation of efficient WAM code. Moreover, it would even be easier to implement as a system relation, since most of the relations for which unidirectionality is useful or necessary are predefined. In any case, we make Unidirectional axiomatic since the relation in question would have to be recompiled every time such an assertion went in or out (unless we had some kind of redundant compilation).

Note that *Unidirectional* is a single-valued relation. That is, for a given relation there can be only one set of "input" arguments specified.

## 5.3 Predefined Relations

A set of relations are already defined when RAD is started. Some of these predefined relations are essential to the operation of RAD, while others are predefined strictly for the convenience of users.

There are three important properties that the relations described below can have:

- Action Some relations are action relations. Only a proposition beginning with an action relation can appear as a consequence of a forward rule. Intuitively, action relations are those that have some side-effect. Action relations are noted as such when they are described.
- Library Most of the relations defined below are built-in: they are already defined when you start RAD, and they persist after you clear the RAD knowledge base. Some relations are not built-in but are instead defined in RAD library files that must be loaded before the relation is defined. Library relations are noted as such in the descriptions that follow, along with the library file where they are defined.
- Triggerability Most predefined relations form propositions that will not trigger a forward rule. Those that will are noted as such.

## 5.3.1 Type Relations

There are some relations that test the type of their arguments, succeeding if the argument is of the correct type, or failing (and backtracking) if it is not. Note that if the argument is a variable, the variable is "dereferenced" to its binding before the type checking is done.

#### Atom thing

Atom succeeds if thing is a symbol.

#### Atomic thing

Atomic succeeds if thing is either a symbol or a number.

#### Ground thing

Ground succeeds if thing is either atomic, or a cons whose car and cdr are ground.

#### Integerp thing

Integerp succeeds if thing is an integer.

#### Nonvar thing

Nonvar succeeds if thing is not an unbound logical variable.

### Numberp thing

Numberp succeeds if thing is a number.

## Var thing

Var succeeds if thing is an unbound logical variable.

## 5.3.2 Comparison Relations

These three relations compare their arguments.

## Eq thing1 thing2

Eq succeeds if thing1 and thing2 are exactly the same object, either:

- · Both lisp objects and EQ, according to lisp, or
- The same unbound logical variable.

## Neq thing1 thing2

Neq succeeds if thing1 and thing2 are not exactly the same object, i.e., it succeeds exactly when Eq fails and vice versa.

= thing1 thing2

= succeeds if thing1 and thing2 unify.

#### 5.3.3 Metarelations

The following relations take one or more propositions as arguments. They can be used in the antecedents of forward rules or backward rules.

And prop1 prop2

Both arguments are clausal

And succeeds if both prop1 and prop2 are provable, with compatible bindings for the logical variables. Since there is an implicit And around the antecedents of a rule, this explicit And is really only useful inside an Orr. And and Orr can be arbitrarily nested.

Bagof template proposition bag

proposition is clausal

Bagof finds all the proofs of proposition. For each proof, Bagof collects an instance of template which is consistent with proposition, and finally unifies this list with bag.

Excuse proposition

proposition is clausal and triggerable

Excuse succeeds if proposition is provable via backward chaining. Excuse prevents any proof of proposition from affecting the justification of the proof in progress. Thus, no JTMS record of proposition will appear in any result of a proof of Excuse.

Known proposition

proposition is clausal and triggerable

Known succeeds if proposition is provable via relation ground assertions, i.e., without resorting to rules, nonground assertions, or slot values.

Orr prop1 prop2

Both arguments are clausal

Orr succeeds if either prop1 or prop2 is provable. Note that RAD tries the two propositions in order: successive proof attempts of Orr will cause successive proof attempts of prop1 until failure. Then prop2 is tried.

Provable proposition

proposition is clausal

Provable succeeds if proposition is provable using backward chaining—from the point of view of backward chaining the proposition (provable (foo ?x)) is identical to (foo ?x). Provable is useful in antecedents of forward rules because it is not triggerable.

Prove agent-name proposition

proposition is clausal

Prove succeeds if proposition is provable by the agent agent-name. The agent returns all possible instantiations of proposition, which are then used one at a time. For each such clause, an assertion that the remote agent SAID that clause is recorded.

Prove-once agent-name proposition

proposition is clausal

Prove-once succeeds if proposition is provable by the agent agent-name. If so, an assertion is recorded that the remote agent SAID that instance of the proposition.

Query agent-name proposition

proposition is clausal

Query succeeds if proposition is provable by the agent agent-name. The agent returns all possible instantiations of proposition, which are then used one at a time. For each such clause, an assertion that the remote agent SAID that clause is recorded. Unlike Prove, the remote agent does not remember the proofs it generates.

Query-once agent-name proposition

proposition is clausal

Query-once succeeds if proposition is provable by the agent agent-name. If so, an assertion that the remote agent SAID that instance of the proposition is recorded. Unlike Prove-once, the remote agent does not remember the proof it generates.

Remember proposition

proposition is clausal

Remember succeeds if proposition is provable. If proposition is proved using a backward rule or some assertion more general than proposition, then proposition is asserted with the proper justification as a side-effect of the Remember proof. Future attempts to prove proposition will not have to reprove it using the same method.

Said agent-name proposition

proposition is clausal

Said succeeds if the agent agent-name said proposition to this agent.

Unless proposition

proposition is clausal

Unless succeeds if proposition is unprovable, much like Prolog's not. Unlike not, it will create a nonmonotonic dependency between proposition and the results (if any) of the proof of Unless. (Unless is actually not implemented as a relation, but rather as a macro for the RAD reader. However, for all practical purposes, it can be treated as a relation.)

Was-told agent-name proposition

proposition is clausal

Was-told succeeds if this agent was told proposition by agent agent-name.

## 5.3.4 Pathological Relations

Fail

The relation Fail always fails and backtracks.

True

The relation True always succeeds.

## 5.3.5 RAD Knowledge Base Relations

The following relations affect the state of the RAD knowledge base, and hence they are all action relations suitable for consequents of forward rules. They are also sometimes useful as antecedents of backward or forward rules. If used in that manner, they always succeed, and in succeeding change the RAD knowledge base in the manner specified for each relation.

Accept proposition

Action; proposition is clausal

Accept causes proposition to be asserted in the RAD knowledge base. The justification of proposition is that the agent who made the accept "said so" and that the agent is reliable unless known to be unreliable.

Assert proposition

Action; proposition is clausal

Assert causes proposition to be asserted in the RAD knowledge base. The justification of proposition is that of the proof so far, either the proof-in-progress for an Assert that appears as a rule antecedent, or the complete proof of the forward rule for an Assert that appears as a forward rule consequent.

Assert-Instance class frames

Action

Assert-instance creates instances of the class class, with names given by frames. When it is used as a consequent of a forward rule, this action relation enables a class system to be expanded at run-time.

Assert-Subclass class frames

Action

Assert-subclass creates subclass relationships from the superclass class to each of its subclasses given by frames. When it is used as a consequent of a forward rule, this action relation enables a class system to be refined at run-time.

54

Consult pathname

Action

Consult loads a RAD file into the knowledge base. If the file has already been loaded, consult will not reload it. RAD consult files are described in more detail in Chapter 9.

**Contradiction** Action

Contradiction creates a contradiction and thus invokes the contradiction resolution mechanism (see Chapter 8).

Erase proposition

Action; proposition is clausal

Erase causes proposition to be OUT. The justification of the erasure is that of the proof so far, as with Assert.

Kill proposition

Action; proposition is clausal

Kill deletes all evidence of proposition from the RAD knowledge base.

Remove-Subclass class frames

Action

Remove-subclass deletes existing subclass relationships from the superclass class to each of its subclasses given by frames. When it is used as a consequent of a forward rule, this action relation enables a class system to be modified at run-time.

Tell agent proposition

Action; proposition is clausal

Tell causes proposition to be asserted in the knowledge base of the agent agent, as described in Chapter 2. The justification for proposition has an IN-list consisting of reliable (SOURCE-AGENT) and said (proposition SOURCE-AGENT, and an empty OUT-list, where SOURCE-AGENT is the name of the agent that executed the relation Tell. The SOURCE-AGENT maintains a justification consisting of either the proof-in-progress for a Tell that appears as a rule antecedent, or the complete proof of the forward rule for a Tell that appears as a forward rule consequent.

## 5.3.6 Class System Relations

The following relations are used for testing or generating classes and instances in the RAD class system. Each of the relations is multidirectional. For instance, one can use subclass to test whether GRADER is a direct subclass of EMPLOYEE (by querying subclass(grader employee)), to find classes that are direct subclasses of EMPLOYEE (via successive proofs of subclass(?x employee)), or to find pairs of classes such that one is a direct subclass of the other (using subclass(?x ?y)).

**Instance** frame class

Library: builtins.rad

Instance is true if frame is an instance of class, that is if the type of frame is class.

Instance\* frame class

Library: builtins.rad

Instance\* is true if frame is an instance of class, that is if the type of frame is a Subclass\* of class, as described in Chapter 4.

Subclass sub super

Library: builtins.rad

Subclass is true if sub is a direct subclass of super. Subclass is the primitive class relation built from the command assert-subclass, as described in Chapter 9.

Subclass\* sub super

Library: builtins.rad

Subclass\* is true if sub is a subclass of super. Subclass\* is the reflexive transitive closure of Subclass, as explained in Chapter 4.

## 5.3.7 Interface with Lisp

RAD has a fast and versatile interface with lisp, available through the following relations.

**Is** thing lisp-form

Action

Is evaluates the lisp s-expression lisp-form, with the following caveats:

- 1. Lisp special forms (e.g., setq) cannot be used in *lisp-form*. (Instead of setq, use set.)
- 2. Evaluation of a RAD logical variable is interpreted as dereferencing. Thus if lisp-form is (cons ?x a), ?x is bound to a, and a is bound to 5, Is will evaluate the form to produce a cons of the symbol a and the number 5.

The result of evaluating the form is then unified with thing. If unification is unsuccessful, Is fails.

It is anticipated that most uses of *Is* will fall into a few stereotypical cases. Therefore a more concise syntax has been implemented within the *RAD* reader. Escapes to Lisp can be prefixed by a comma; the *RAD* reader will parse these and generate the appropriate calls to *Is*. (Refer to Chapter 9.

Action Do lisp-form

Do evaluates the lisp s-expression lisp-form. The evaluation caveats of Is also apply to Do. Do always succeeds, so to be useful, lisp-form should perform some side-effect.

## Querying the User

prop is clausal; Library: ask.rad Ask prop Ask takes a proposition as an argument and queries the user for an instance of this proposition (with exceptional cases described below). If the user replies positively with a valid instance, then that instance is automatically inserted in the knowledge base and Ask then succeeds by matching the goal with the new assertion. If not, then Ask fails.

For example, if information pertaining to a relation P may only be supplied directly by the user, then P may be defined by a single rule:

Infer P(?X ?Y) from Ask(P(?X ?Y))

If Q is defined by

Infer Q(?X ?Y) from P(?X ?Y)

then

User: ? q(a ?x) Zeus: P(A ?X)?

User: reply zeus ''p(a 1)''

Zeus: Assertion P(A 1) accepted.

User: ? q(b ?X) Zeus: P(B ?X)?

User: reply zeus '','

Zeus: NO SOLUTION

Here, the user responds positively to the first query with the instance P(A 1), and negatively to the second with an empty string.

Note that a query may be repeated, as it may have several valid responses:

User: ? q(a ?x)
Zeus: P(A ?X)?
User: reply zeus ''p(a 1)''
Zeus: Q(A 1)

User: next
Zeus: P(A ?X)?
User: reply zeus ''p(a 2)''
Zeus: Q(A 2)

In order to avoid unnecessary repetitions of queries, the system maintains a history of the queries that have already been posed to the user. If the user has already responded negatively to a query about a given goal (or any goal of which the given goal is an instance), then the query is cancelled and Ask simply fails:

User: ? q(?x ?y)
Zeus: P(?X ?Y)?
User: reply zeus ''''
Zeus: NO SOLUTION
User: ? q(a ?X))

Zeus: NO SOLUTION

In Chapter 9, we describe the top-level command ask, which allows the system builder to customize the format in which a query is posed.

Ask-once prop

prop is clausal; Library: ask.rad

This relation has the same behavior as Ask, except that once the user has been queried about a goal, Ask-once will never query him about it again (regardless of the user's original response). This is useful in case it is known that a goal has only one valid instance:

User: Infer age(?X ?Y) from ask-once(age(?X ?Y)))
Zeus: Backward Rule AGE-1 accepted.

User: ? age(bob ?X)
Zeus: AGE(BOB ?X)?

User: reply zeus ''age(bob 35)''

Zeus: AGE(BOB 35)

User: next

Zeus: NO SOLUTION

### 5.3.9 Miscellaneous Built-In Relations

Count-proofs proposition count Proposition is clausal; Library: misc.rad Count-proofs is true if there are exactly count proofs of proposition in the RAD knowledge base. Note that the JTMS status of count-proofs proof is weird: it does not depend on the status of any of the actual proofs counted.

Element elt list

Library: misc.rad

Element is true if elt is an element of the list list. (This is the same as member from Prolog.)

Frules-indexed reln rules

Library: misc.rad; System Axiomatic

Frules-indexed relates a relation reln to the forward rules in which it appears as an antecedent. It is *Unidirectional*, i.e., its first argument must be ground.

Host database name

Library: dai.rad

The relation *Host* is defined for the built-in class **DATABASE**. Its value, supplied by *RAD*, is the name of the host machine for the database.

Print format print-elements

Library: misc.rad

Attempting to prove *Print* or asserting *Print* as a consequent of a forward rule will cause output to be printed to the window of the user interface process. *format* is lisp's format directive string, while print-elements is a list of arguments for that string.

Reliable agent

Library: dai.rad

The relation Reliable is used to provide justifications for data learned from other agents.

Unreliable agent

Library: dai.rad

The relation *Unreliable* is used to provide justifications for data learned from other agents.

!

RAD supports the prolog control primitive cut ("!"), both in backward rules, and as described in Chapter 7 in forward rules as well. The operational semantics of cut is identical to that in Prolog, and beyond the scope of this manual.

### 5.4 User-Defined Relations

Corresponding to each user-defined class is a (possibly empty) set of user-defined relations associated with that class. Every relation can be defined for a unique class by means of the relation Argument-types (defined in Section 5.2). The frames that may have values for a given relation are the instances of the class for which it is defined, i.e., the instances of the class listed first in Argument-types. Thus, in the example of Figure 4.2, if relations called Hourly-wage and Title are defined for the classes EMPLOYEE and FAC-ULTY, respectively, then the frames that may have Hourly-wage values are NAT, HILARY, ANN, and DON, while only ANN and DON may assume Titles.

The manner in which relations are inherited from their defining classes provides motivation for the construction of user-defined metaclasses, such as those of Figure 4.3. If ANIMALIA, CHORDATA, etc. had simply been defined as instances of CLASS, there would have been no way for them to acquire relation values. But as instances of the user-defined class BIOLOGICAL-CLASS, they may assume values for any relations defined for that class. If Common-name is among these relations, then the statement "the Common-name of PROTOZOA is ONE-CELLED-ANIMAL" makes sense. The partitioning of BIOLOGICAL-CLASS into subclasses allows the definition of relations that pertain to some biological-classes but not to all. For example, one might refer to the number of species of protozoa, but not to the number of species of amoeba, while the number of species of animalia is not practically measurable. The relation Number-of-species, therefore, should not be defined for the class BIOLOGICAL-CLASS, but rather for its subclass PHYLUM.

		8

# Chapter 6

# **Backward Inference**

The primary means by which computation is carried out in RAD is through a goal-directed theorem-proving process called backward inference. This process permits goals, in the form of propositions, to be proven by a combination of unification and backward chaining.

When a proposition is presented to the RAD theorem prover as a goal, it attempts to derive an instance of it from the data in the knowledge base. An instance of a proposition is a second proposition that results from the first by performing some set of variable substitutions.

One way in which it might do this is to unify the goal with an assertion. The process of unification amounts to finding the most general common instance of two propositions. If a goal is unifiable with an assertion that is IN, then the resulting instance of the goal is returned as the result of the proof. For example, if the knowledge base of Figure 4.2 contains

#### DESCRIPTION (?X:STUDENT IDEALISTIC)

then the goal DESCRIPTION(?X:EMPLOYEE ?Y) could succeed by returning the instance DESCRIPTION(?X:(GRADER TA) IDEALISTIC).

A goal may also be proved with the use of a backward rule. A backward rule is composed of a proposition, called its consequent, and one or more antecedents. A rule represents the belief that any instance of its consequent is true whenever any compatible instances of its antecedents are true. If a goal is unified with the consequent, then the corresponding instances of the antecedents become subgoals—recursively proving all of these subgoals completes the proof of the original goal. This process is known

as backward chaining or goal-directed inference, and is the basis of Prolog [Clocksin and Mellish 1981] and other logic programming systems.

For example, in order to derive a value for MICHAEL's *Uncle* slot, the goal UNCLE(MICHAEL ?X) may be unified with the consequent of the rule

creating the subgoals PARENT (MICHAEL ?Z) and BROTHER (?Z ?Y). Suppose that the first of these is matched with the consequent of

Parent-rule: Infer parent(?X ?Y) from mother(?X ?Y)

and is thus replaced by the subgoal MOTHER (MICHAEL ?Y), which is matched with

Instance slot value MOTHER-7
 MOTHER(MICHAEL SUZY)

The second subgoal, which becomes BROTHER (SUZY ?Y), may then be derived from

Instance slot value BROTHER-23 BROTHER(SUZY DAVID)

The instance UNCLE (MICHAEL DAVID) of the original goal is thereby proved by backward chaining.

The RAD backward inference mechanism is based on an extended Warren Abstract Machine emulator [Warren 1983]. The assertions and backward rules associated with a given relation, which form its definition with respect to backward inference, are compiled into a sequence of instructions that are interpreted by the emulator when a goal is processed. Although the user need not be aware of the workings of the compiler and the emulator, the code generated for a relation may be examined by means of the top-level command bgrind (Chapter 9).

Certain built-in relations (see Chapter 9) are defined directly by WAM instructions, either for efficiency or because their desired behavior cannot be represented by rules. An example of a relation that could not effectively be defined in terms of rules is the built-in relation *Unless*. This relation takes a single argument, which is a proposition, as in

When an antecedent of this type is processed, the system attempts to prove the proposition that occurs as the argument of *Unless* (under the current variable bindings). If this proof attempt fails, then the subgoal succeeds; if the proof succeeds, then the subgoal fails.

# 6.1 Asserting Proof Results

When a proposition proved by backward chaining is explicitly added to the knowledge base as an assertion, it receives a justification that is constructed upon examination of the proof. Thus, the proposition derived in the first example of this chapter results in

Instance Slot Value UNCLE-2
 UNCLE(MICHAEL DAVID)

which acquires a justification with IN-list (MOTHER-7 BROTHER-23 UNCLE-RULE PARENT-RULE), i.e., all the data involved in the proof, and OUT-list ().

Nonmonotonic dependencies are constructed from proofs that involve *Unless* antecedents. For example, if the proposition (HAS-CHILD SUZY) were derived from the backward rule Parent-rule-2 and Assertion MOTHER-7 above, and

Assertion HAS-CHILD-5 HAS-CHILD(SUZY)

were derived as a result, then the IN-list of its justification would be (MOTHER-7 PARENT-RULE-2), but the justification would also reflect the dependency of the derivation on the failure to prove ADULT(MICHAEL). This would be done by making the OUT-list (ADULT-2), where

Assertion ADULT-2 (OUT)
ADULT(MICHAEL)

is an unjustified assertion, created for the purpose of this justification (unless it already existed). If ADULT(MICHAEL) were to be asserted later, then ADULT-2 would become IN, and HAS-CHILD-5 would go OUT, as it should.

The case of an UNLESS goal with unbound variables presents a new problem. Suppose, for example, that we have a relation *Orphan* with an associated rule

Then the goal ORPHAN(ANNIE) will succeed if PARENT(ANNIE ?Z) fails. In this case, the new justification for

Assertion ORPHAN-2: ORPHAN(ANNIE)

should contain only ORPHAN-RULE in its IN-list, but there is no assertion, general or particular, which could be placed in the OUT-list to record the nonmonotonic dependency. This problem is solved by the introduction of a new datatype:

Failed Goal PARENT-2 (OUT)
PARENT(ANNIE ?Z)

A failed goal is a datum that is created only in this situation. When a proof succeeds as a result of a failure to prove a proposition that follows *Unless* in an antecedent, the proposition that failed is inserted in the knowledge base without justification as a failed goal. It represents the belief that some instance of the proposition is true. That is, any variables in a failed goal are understood to be existentially quantified. The failed goal PARENT-2 above, which represents the belief that ANNIE has some parent, would appear in the OUT-list of the justification of ORPHAN-2.

If some instance of a failed goal is asserted at any time, the system automatically creates a monotonic dependency of the failed goal on the assertion. Thus, if the assertion ORPHAN-2 were justified as described above, and

#### Assertion PARENT-3:

PARENT(ANNIE WARBUCKS)

were later to become IN, then the failed goal PARENT-2 would also be forced IN and hence ORPHAN-2 would go OUT.

# 6.2 Proof Strategies

In the process of proving a goal by backward chaining, the goal is matched against each element of a set of assertions and rules until unification succeeds. If a resulting subgoal fails and the prover backtracks to the original goal, the matching process is resumed until another match is found.

#### 6.2.1 Relations

It is not surprising that the ordering of the list of data to be matched against a goal may not only affect the efficiency of a proof but may actually determine whether a proposition is provable at all. Consider, for example, the relation *Not-instance\**, defined by

```
Infer not-instance*(?X ?C:class)
from (instance*(?X ?C) ! fail())
and
not-instance*(?X ?C)
```

(The symbol "!" is the standard Prolog cut and Fail is a built-in relation that represents the empty relation.) This relation, which represents X is not an instance\* of class C, behaves as desired only if the data are applied in the order indicated, with the rule preceding the assertion.

In Prolog, the order in which a set of data is created is strictly preserved by the inference system. Of course, this means that a relation must be recompiled whenever a new rule or assertion is added. One difference between an expert system built with RAD and most Prolog programs is that an expert system is likely to generate a large number of assertions dynamically during its execution. In order to avoid excessive recompilation, RAD makes a distinction between ground assertions (i.e., assertions without variables) and rules. When a ground assertion is created, it is not compiled, but instead inserted at the end of a list that is interpreted directly by the inference mechanism. In an attempt to prove a goal, the list of relevant ground assertions is always processed first, followed by the (compiled code for the) relevant rules. Nonground assertions—like backward rules—are compiled in the order in which they are asserted.

## 6.2.2 Multiple-Valued Relations

Different goal-matching strategies are required for relations associated with frames, i.e., specificity-ordered relations. For these, each instance slot value is attached to the frame to which it pertains. In fact, this is an important advantage of declaring a relation to be specificity-ordered: when a goal is being processed that concerns values for a fixed frame, the values for that frame only are retrieved, while other frames are ignored.

In the case of a multiple-valued relation, all relevant instance slot values are matched against a goal before any backward rules are applied. The first argument of a goal pertaining to a slot may be either a variable or a frame. If it is a frame, then the instance slot values for that frame are tested first, followed by the class slot values for all superclasses of the frame's type (in depth first order). Finally, a list of all the backward rules for the slot is traversed.

If the first argument of the goal is a variable, the strategy is complicated. In this case, each class that has a subtype in common with the type of the variable must be examined for matching values. Consider, for example, the goal (DESCRIPTION ?X:STAFF ?Y), where Description is the multiple-valued relation that was defined for class PERSON in Chapter 4. In processing this goal, the g.l.b. of STAFF and each subclass C of PERSON must be computed. There are three possibilities:

- 1. The g.l.b. of STAFF and C may be BOTTOM. (This occurs when C is FACULTY.) In this case, C is ignored.
- 2. The g.l.b. may be C itself, i.e., C may be a subclass of STAFF. (This occurs when C is STAFF, GRADER, or TA.) In this case, the variable ?X may be bound to either an instance of C or to a variable whose type is C. The list of all instances of C is examined first. Each instance becomes the binding of ?X while ?Y is bound to each of its Description values. After all of these instance values have been examined, the type of ?X is coerced to the g.l.b. and the class values of C are examined in order.
- 3. The g.l.b. may be a nontrivial proper subtype of C. (This occurs when C is PERSON, STUDENT, GRADUATE, or UNDERGRAD-UATE.) In this case, ?X may be bound to he instances of C, which

are therefore ignored, but the class values of C are examined as in case 2.

After all classes and all relevant slot values have been examined, the list of backward rules is traversed as in the previous case.

### 6.2.3 Single-valued Relations

The strategy for satisfying a goal pertaining to a single-valued relation is somewhat simplified by the design decision that a query concerning the values of a single-valued relation for a fixed frame should not return more than one instance. Therefore, in the case in which the first argument of the goal is a frame rather than a variable, only one instance of the goal may be proved—there is no backtracking. To satisfy such a goal, the frame is first examined for an instance value that is IN. (There can be only one such value.) If none is found, then the system tries to derive a value using backward rules. Only if this fails is the default value for the appropriate class retried. Thus, default values, unlike class slot values for multiple-valued relations, are used only as a last resort.

If the first argument of the goal is a variable, then the prover begins by binding this variable to some instance of its type, and attempts to satisfy the goal according to the procedure described in the preceding paragraph. If this fails, or if an additional solution is sought, then the variable is bound to another frame, until it has been bound to all possible frames. This accounts for the result of the query in Chapter 4 for all provable instances of (NATIONALITY ?X:PERSON ?Y). Note, in particular, that although the proposition (NATIONALITY ?X:STUDENT CHINESE) has been asserted as a default value, and this value remains IN, this proposition is not retrieved as a provable instance of the goal. The reason for this is the difference between default values of single-valued relations and class values of multiple-valued relations—a default value for a class is not necessarily inherited by the instances of the class. Therefore, the proposition (NATIONALITY ?X:STUDENT CHINESE) should not be considered a general assertion. A default value should be used only in computing a value for a particular object, and not for a class of objects.

# Chapter 7

# Forward Inference

A backward rule has effect only when it is relevant to a goal being processed by the system. The insertion of a backward rule, therefore, affects only the implicit informational content of the knowledge base, without causing new assertions to be added explicitly. Consider, for example, the backward rule PARENT-RULE of Chapter 6, which states that all mothers are parents.

Parent-rule: Infer parent(?X ?Y) from mother(?X ?Y)

In the presence of the assertion

Instance slot value MOTHER-7
MOTHER(MICHAEL SUZY)

this rule enlarges the implicit knowledge base to include the proposition PARENT(SUZY MICHAEL), without actually creating a new assertion.

The same logical implication expressed by PARENT-RULE could alternatively be represented as the *forward rule*:

Parent-rule-3: If mother(?X ?Y) then parent(?X ?Y)

While the two rules are logically equivalent, they are used quite differently. The forward rule takes effect not when a goal matches its consequent PARENT(?X ?Y), but rather when an assertion matches its antecedent MOTHER(?X ?Y). In this event (assuming the new assertion is IN), another

assertion, representing the corresponding instance of PARENT(?X ?Y), is automatically added to the knowledge base. The PARENT assertion is then justified by the MOTHER assertion and the rule. Thus, when the match between MOTHER-7 and PARENT-RULE-3 is discovered, the result is the new datum

Instance Slot Value PARENT-4
PARENT(MICHAEL SUZY)

which is justified with an IN-list (MOTHER-7 PARENT-RULE-3) and OUT-list NIL. This process is known as forward chaining or data-directed inference.

## 7.1 Overview of Forward Chaining

A forward rule may have any number of antecedents and consequents. Antecedents have the same form as those of backward rules. When a new assertion is unified with an antecedent of a forward rule, the set of remaining antecedents is presented to the backward inference system as goals. For each simultaneous proof of these goals, a *firing* of the rule occurs, i.e., its consequents are processed.

Consequents of a forward rule are actions and involve one of the action relations, such as Assert, Consult, Do, Erase, Is, Kill, Print, or Tell. Intuitively, these are all relations that have some side effect—either to the RAD knowledge base, to another agent, to the user interface, or to the Lisp interpreter. All consequents that are not explicitly actions are considered to be Assert actions. Thus, if if you want to assert some proposition as a consequent of a forward rule, you need only write the proposition itself; the Assert relation is implicit.

When a rule is fired, each of the consequences is executed. For assertions, execution means that an instance of the proposition to be asserted is actually asserted. The justification for this assertion is constructed from the data involved in the derivation of the antecedents, as described in Chapter 6. For other sorts of action propositions, the corresponding action is executed. The actions are described with the rest of the built-in predicates in Chapter 9.

Suppose, for example, that the knowledge base contains

```
then (under-treatment(?X) print("Prescription for ~A: ~A" (?X ?Y)))
```

when

Assertion PATIENT-2 PATIENT(BILL)

is added. The first antecedent of PATIENT-2 is matched with RULE-1, triggering an attempt to prove SHOULD-TAKE(BILL ?Y). Suppose that the instance SHOULD-TAKE(BILL ASPIRIN) is derived. Then after

Assertion SHOULD-TAKE-1 SHOULD-TAKE(BILL ASPIRIN)

is added to the knowledge base, the rule RULE-1 fires:

Assertion UNDER-TREATMENT-1 UNDER-TREATMENT(BILL)

is added, justified by RULE-1, PATIENT-2, and SHOULD-TAKE-1, and

Prescription for BILL: ASPIRIN

is printed.

Some thought is required in determining whether a given implication should be represented as a forward rule or a backward rule. A backward rule offers the advantage of increasing the inherent knowledge of a system without incurring the expense (in both time and space) of creating new assertions. It may be necessary, however, for this knowledge to be represented explicitly in order for it to take some desired effect.

Suppose, for example, that the assertion

Assertion ORPHAN-3 (IN)
ORPHAN(GEORGE)

is added as a result of the rule ORPHAN-1 of Chapter 6:

 Then the OUT-list of its justification contains a datum corresponding to the last antecedent of the rule,

Failed Goal PARENT-5 (OUT)
PARENT(GEORGE ?Z)

which was unprovable at the time ORPHAN-3 was created. Suppose that

Assertion MOTHER-12 MOTHER(GEORGE MARY)

were asserted later. It would then be desirable for PARENT-5 to come IN and for ORPHAN-3 to go OUT. The backward rule PARENT-1, however, could not cause this to occur. Although an instance of PARENT-5 would become provable, that instance would not be discovered. It would probably be preferable in this case to code the rule in the form of the forward rule PARENT-RULE-3 instead. This would produce a new assertion

Assertion PARENT-6
PARENT(GEORGE MARY)

on which PARENT-5 would become monotonically dependent, and ORPHAN-3 would go OUT as desired.

# 7.2 Forward Chaining in More Detail

The above description of forward chaining is accurate but not very complete. For some RAD programs, it is important to know not only what happens but exactly when things happen.

When a unit clause is asserted, the new assertion is placed on an agenda. As long as the agenda is nonempty, RAD does not return to the user. Instead elements are popped from the agenda and executed—used to trigger forward rules. (By default the agenda is LIFO—a stack—but a user can control which element is picked from the agenda with the function pick-cselt.)

When an agenda element is picked, RAD looks for an antecedent from a forward rule to unify with it. If more than one such antecedent will unify, the antecedent from the rule with the highest priority is chosen. Priorities are integers from 1-99 associated with forward rules, and a rule has priority 50 by default. A forward rule with a different priority is created by enclosing a two-digit number in square brackets before the If clause of the rule:

When an antecedent from a rule has been chosen, RAD attempts to unify the agenda element with the antecedent. If the unification fails, RAD looks for another antecedent, choosing antecedent from rules with higher priorities over antecedents from rules with lower priorities.

If the unification succeeds, RAD attempts to find proofs of the conjunction of the remaining antecedents in the context of the variables bound in the successful unification. For example, if the agenda element BAR(12 ?X) triggers the above rule, then RAD attempts to prove AND(FOO(?x 12) BAZ(?z?x)). These proofs are found by standard RAD backward chaining as in Chapter 6. In particular, no OUT rules or clauses will be used in any proof of the remaining antecedents.

For each proof of the antecedents, RAD executes the consequences once. This means that a single triggering of a single forward rule can lead to multiple firings of that rule, if there happen to be multiple proofs of the remaining antecedents in the knowledge base. Since the remaining antecedents are just a conjunction RAD propositions, a well placed cut ("!") among the antecedents can served to prune the proofs. For example, the rule

will allow (at most) one firing of the rule for every triggering of the antecedents.

Some of the consequents of the fired rule may assert new facts. These assertions are added to the RAD knowledge base, and if they are unit clauses, they are added to the agenda. Nothing new is picked from the agenda until the currently active agenda element has triggered all the rules it can.

After finishing with one antecedent, RAD looks for another antecedent that will unify with the current agenda element, continuing the process of firing forward rules. Only when there are no more antecedents to unify will another element be picked from the agenda.

Some antecedents trigger their rules by unifying their first argument rather than the whole proposition with the active agenda element. For example, if KNOWN(FOO(?X)) (described in more detail in Chapter 5) is an antecedent in a rule, it can be triggered by FOO(?X:NUMBER). This is consistent with Known's semantics: Known is true if there is a ground IN instance of its first argument in the RAD knowledge base—i.e., if its first argument is provable without resorting to nonground clauses or backward rules. Thus it makes sense for Known to be triggerable by unification of the agenda element with Known's first argument.

Some propositions (e.g., those involving the relation Bagof) cannot be triggered by an agenda element. The special predicates that are not triggerable are noted in Chapter 5.

# Chapter 8

# Contradiction Resolution

### 8.1 Introduction

RAD contradiction resolution performs a restricted sort of abductive inference [Pierce]. A RAD contradiction represents a user-designated conflicting set of beliefs. Given a contradiction, RAD attempts to hypothesize a set of propositions that, if believed, would cause disbelief in at least one element of the conflict set. Further, RAD attempts to create a consistent justification for the hypothesized propositions. When successful in doing so, RAD is said to have resolved the contradiction. In terms of the JTMS, contradictions are resolved by retracting some set of assumptions in the foundations of the JTMS. An assumption is a datum that is currently believed (IN) based on a valid justification with a nonempty OUT-list. An assumption is retracted if an elective in the OUT-list of the assumption receives a valid justification. RAD contradiction resolution involves attempts to disbelieve assumptions and to find reasons to believe electives.

RAD contradictions are instances of a special kind of datum. They may have only a single justification and the JTMS will always attempt to find a stable, well-founded knowledge base state of support-statuses in which all contradictions are OUT. If none can be found, the Contradiction Resolution Mechanism (CRM) will attempt to produce a set of electives and associated new justifications which will make such a state possible. If it is successful, the contradiction is OUT and hence resolved. If the CRM fails, the contradiction is added to a set of unresolved contradictions that remain IN. Resolution of

these is attempted whenever they are consequences of data whose support status is altered by the JTMS. The RAD CRM implements the extended dependency-directed backtracking algorithm described in [Petrie 1987].

The semantics of a RAD contradiction is that it represents an undesirable situation, indicated by a combination of belief of the data in the IN-list and lack of belief in the data in the OUT-list of the justification for the contradiction. Although the term "contradiction" is used for historical reasons, in a justification-based TMS such as used by RAD, contradictions have no logical import and only denote a user-defined problem. There is no requirement in RAD that if an assertion and its negation are both IN, then there is a contradiction. In fact, there is no definition of "negation" in RAD. RAD does not try to prove the logical consistency of its knowledge base. A user must explicitly specify the conditions that RAD will recognize as a contradiction in the knowledge base.

The user can specify these conditions for a contradiction either interactively or via a rule. If a forward rule has the assertion CONTRADICTION() as its consequent, then when that rule fires, a contradiction is inserted into the knowledge base with the same justification as any other consequent of that forward rule. A user can contradict a conjunction of beliefs by using their names as arguments to the top-level command contradict. The last answer supplied by a top-level query can be contradicted by the command wrong.

The relations, Fix, Prefer, and Defeat enable a user to provide domain knowledge to guide contradiction resolution. These relations are specially recognized by the RAD CRM. They represent explicit knowledge about the use of other assertions in an inference procedure, and so represent a metalevel of knowledge. As a result, they are called "metarelations".

## 8.2 The Contradiction Resolution Process

### 8.2.1 The FIX Phase

The CRM is based on a simple notion of defeasible reasoning. As a simplification (see *Defeat* below for a complication), a valid justification is *defeasible* if it has a nonempty OUT-list or if every valid element of the justification set of some element of its IN-list is defeasible. We say that an assertion is defeasible if every valid justification supporting it is defeasible. The classi-

cal definition of an assumption in a JTMS is a special case of a defeasible assertion. This more general notion of defeasibility suggests a recursive procedure. Given any culprit to be defeated, i.e., made OUT, the procedure is to attempt to defeat all of its valid justifications. For each justification, this is done by attempting to make some element of its OUT-list IN (namely an elective) or by taking one of the elements of its IN-list as a new culprit to defeat. Ultimately, the result will be an AND of assertions (electives) to be made IN. Obviously, there will be many possibilities. The purpose of the metarelations is to suggest guidance for this search.

The relation Fix takes three arguments: the target, the fix-culprit, and the fix-elective. A Fix is defined for a particular potential contradiction if the Fix has a target that will unify with an IN supporter of that contradiction should it be created. The purpose of such a Fix is to identify some particular assumption (the fix-culprit) that is likely to be in the foundations of the contradiction and, if so, for which a retraction attempt should be made. The fix-elective identifies a possible elective for that assumption. That is, the Fix may so suggest a possible way of retracting the assumption.

The CRM begins with a datum, called the culprit, to be made OUT. Initially, the culprit is the contradiction. First, an attempt is made to use domain knowledge to make the culprit OUT by using a Fix assertion to invalidate one of the IN supporters of the culprit. Each of the IN supporters in turn becomes a culprit for which there may be a Fix. An available Fix for a culprit is obtained when the culprit proposition unifies with the target of some provable Fix. A qualified Fix is derived from the available Fix when its fix-culprit unifies with the proposition of some datum in the foundations of the culprit (the target of the Fix). If some element of the OUT-list of the supporting justification of the fix-culprit unifies with the fix-elective of the qualified Fix, then the Fix is applicable and that element becomes the elective to be made IN by the CRM (making a datum IN is described in Section 8.2.2).

An example of a Fix assertion might be

```
Fix(Conclusion(?Patient ?Degree ?Meas ?Chem)
    Symptom(?Patient ?Symptom)
    Mistaken-observation(?Patient ?Symptom))
```

<sup>&</sup>lt;sup>1</sup>Any of these arguments may be left unspecified. However, the more specificity, the more useful fixes will be. An assertion of Fix(?X ?Y ?Z) would not be useful.

This Fix asserts that if any conclusion becomes a culprit to be made OUT, and the conclusion depends on belief in some symptom, then a mistaken observation of that symptom may retract belief in it. That mistaken observation will only become an elective if it is either already in the OUT-list of the supporting justification of the symptom, or can be placed there by a Defeat assertion.

A Defeat assertion provides a way of inserting an Unless condition retroactively. An Unless condition in a rule indicates a non-monotonic condition. One way of avoiding performing the computation involved in testing this condition is to use the Known and Unless relations in conjunction. However, this will check to see if the condition is asserted in the database and will invalidate the result of the rule firing if the condition is ever asserted. In contrast, a Defeat assertion represents a condition that will never even be considered unless the result of the rule firing becomes implicated in a contradiction.

The Defeat metarelation also takes three arguments. The first of these is the defeat-culprit, the second is the in-supporter, and the last is the defeat-elective. A Defeat asserts that the defeat-elective may be added to the OUT-list of each justification of the defeat-culprit that contains the in-supporter in its IN-list. Thus the in-supporter is a way of distinguishing among justifications, if needed. If the Defeat is meant to apply to justifications with empty IN-lists (for example, premise justifications), the word none may be given as the in-supporter. If any nonempty IN-list should apply, a variable, such as ?X, can be used for the in-supporter. In general, though, a Defeat represents an argument against a particular reason for believing the defeat-culprit. If the Defeat is meant to apply to a contradiction justification, the defeat-culprit should be the term contradiction. In this case, the in-supporter should unify with one of the antecedents of the rule that produced the contradiction and the defeat-elective will be added to the OUT-list of the justification of the contradiction. An example of a Defeat might be

To illustrate how this *Defeat* and the previous *Fix* could be used, consider a conclusion of a high sodium concentration for a patient named Jane, generated by the following assertion and rules:

Symptom(Jane dehydrated)

If Symptom(?Patient dehydrated)
then Conclusion(?Patient low amt H20)

If (Conclusion(?Patient low amt H20)
 Remember(Normal(?Patient amt Na)))
then Conclusion(?Patient high conc Na)

Infer Normal(?Patient amt Na)
from Unless(Result(?Patient low amt Na))

If the final conclusion of a high sodium concentration for Jane were to support a contradiction, belief in this conclusion could be retracted by hypothesizing that a mistaken observation was made about the symptom which supported this conclusion. The Fix focuses on this hypothesis and the Defeat makes it applicable. Notice that none of the rules includes an Unless clause concerning mistaken observations.<sup>2</sup> The Defeat assertion allows the assertion Mistaken-observation (Jane dehydrated)) to be introduced into the knowledge base and into the OUT-list of the assertion of the symptom during contradiction resolution. This still does not resolve the contradiction. It still remains to make the new assertion IN to resolve the contradiction (described in Section 8.2.2).

There may be more than one Fix that is applicable to a culprit. One of the uses of the metarelation Prefer is to allow the user to specify domain knowledge to decide between Fixes. Available Fixes are ranked on the basis of their fix-electives. An alternative would be to consider the fix-culprits. But the reasoning used is that, ultimately, reasons for distinguishing between assumptions rest on reasons for belief or lack of belief in their electives. It is the electives that compete for valid justifications from the contradiction resolution process. Thus, the electives should be ranked with respect to each other in terms of preference of attempted justification. RAD contradiction resolution uses a very general mechanism of preferences to do so.

<sup>&</sup>lt;sup>2</sup>The use of the *Unless* clause in these rules is worth explaining. The second forward rule has an antecedent that asserts that the patient's amount of sodium is normal. This is proven via the backward rule due to the lack of any reason to believe sodium level is low. By using a backward rule and the relation *Remember*, an assumption of normality is introduced into the database. This is the standard way to achieve a "positive" assumption in a *RAD* database.

One datum is preferred to a second if *Prefer* is the relation symbol of a provable assertion for which the first and second arguments unify with the first and second data, respectively. A preference is such a provable assertion. The semantics is that belief in the first argument is preferred to belief in the second for the purposes of the belief revision being performed by contradiction resolution. One belief is dominated by another if it is provable that the latter is preferred over the former. The CRM will first choose a *Fix*, the fix-elective of which is not dominated by that of another. As an example, suppose we add the following rules and assertions to those above:

```
If (Lab-test(?Patient low ?Meas ?Chem)
    Conclusion(?Patient high ?Meas ?Chem))
then contradiction()

Lab-test(Jane low conc na)

Fix(Lab-test(?Patient ?Degree ?Meas ?Chem)
    Lab-test(?Patient ?Degree ?Meas ?Chem)
    Test-error(?Patient ?Degree ?Meas ?Chem))

Defeat(Lab-test(?Patient ?Degree ?Meas ?Chem)
    none
    Test-error(?Patient ?Degree ?Meas ?Chem))
```

The first rule and assertion will cause a contradiction since the symptoms indicate that Jane should have a high concentration of sodium, but the lab-test shows just the reverse. Now two Fixes are applicable in this case. A doctor will generally recheck the symptoms before he rechecks the test because it is cheap and easy to do so. This domain knowledge can be represented by the assertion

which asserts that belief in a mistaken observation is preferable to that in a lab test error for the purpose of contradiction resolution. Notice also that this

preference could have been provable. For instance, this preference could rest on a default assumption that the patient is not being remotely diagnosed, in which case it is conceivable that rechecking the symptoms might be expensive. Such knowledge could be represented by substituting the following more specific rule for the above preference:

It is sufficiently important to mention again that *Fixes* and *Defeats* can be similarly qualified by the antecedents of rules. Also, no single argument of any of the metarelations need be even partially instantiated. For instance, the assertion

fix(conclusion(?X ?Y ?Z ?U) symptom(?V ?W) ?T)

would cause an attempt to retract any supporting assertion of a symptom with any elective possible. Finally, we note that the same preferences used to choose the order in which Fixes are tried may also be used to guide the dependency-directed backtracking of [Petrie 1987] as explained in the next section.

The Fix operation may be summarized as follows. When a contradiction becomes IN, before dependency-directed backtracking is used to find an elective, the CRM will look for a Fix. The CRM will find all of the available Fixes for the IN supporters of the contradiction. If there are none, then dependency-directed backtracking is used as described below. Given a set of available Fixes, a best Fix will be chosen such that its elective is not dominated by that of another applicable Fix. In the above example, the mistaken-observation Fix would be tried prior to trial of the test-error Fix, since the latter is dominated by the former.

When a Fix is chosen for a culprit, an attempt is made to qualify the Fix and make it applicable. This entails a search of the foundations of the culprit. First, the culprit itself is examined to see if it unifies with the applicable fix-culprit. This is why the first two arguments of a Fix may be identical. Failing a qualification at this level, the IN supporters of the culprit are examined. An attempt is made to find new Fixes which are applicable to these supporters. A new best Fix is chosen at this time. The target of the best Fix becomes

the culprit. In the case that an IN supporter becomes the new culprit, the search begins at that datum. Otherwise, the search continues where it was interrupted by the attempt to collect new Fixes.

Given that the qualification search proceeds to the IN supporters of some datum, an attempt is made to qualify the best Fix for the culprit on each of the supporters. If the Fix becomes qualified, then an attempt is made to make the Fix applicable and make IN the resulting fix-elective. If the qualified Fix is not applicable or the fix-elective cannot be made IN, then the Fix is removed from further consideration and another best Fix is chosen. A Fix is also discarded and a new best Fix chosen when the foundation search of a culprit is complete and the Fix has not been successful.

### 8.2.2 Dependency-Directed Backtracking

If no Fix can be used to make a culprit OUT, dependency-directed backtracking is invoked. Given a culprit to make OUT via DDB, the CRM attempts to either determine that it is already invalid or invalidate each of its justifications. A justification is invalidated by making some element (elective) of its OUT-list IN or (culprit) of its IN-list OUT. The former is tried first. Preferences are used to select the elective or culprit to pursue first. An elective is chosen that is not dominated by any other element in the OUT-list. A culprit is chosen that is not preferable to any other in the IN-list. If the CRM is in interactive mode, the user will be queried to confirm this choice. The candidate electives or culprits will be enumerated. The user may respond with a carriage return only if the first choice is acceptable. Otherwise, one of the other candidates may be selected by supplying its number.

An elective is IN already if it has already been made IN or currently has a support status of IN and is not in the transitive closure of the consequences of data for which the CRM has constructed new consequences. If an elective is not already IN, the CRM will first attempt to make IN an elective by proving it. If the proof is successful, a justification for the elective is provided from the proof. If the elective has a relation symbol for which a query schemata has been defined, the user will be queried as to the belief in the elective. An answer of "yes" gives the elective a premise justification. An answer of "no" indicates that the datum is not suitable as an elective. An answer of "maybe" causes the elective to be abductively justified as described in [Petrie 1987]. This new justification has the semantics that belief in the elective is assumed

in order to resolve the contradiction.

The elective will also receive such a justification if it is provable that belief in the elective is preferable to belief in the contradiction; the first argument of such a preference unifies with the elective and the second is the relation *Contradiction*. In our example so far, the following preferences would be sufficient to allow the CRM to make IN the various possible electives:

Prefer(Mistaken-observation(?Patient ?Symptom) contradiction())
Prefer(Test-error(?Patient ?Degree ?Meas ?Chem) contradiction())
Prefer(Result(?Patient low amt ?Chem) contradiction())

Abductive justifications are constructed so as to become invalid later if belief in the elective is no longer necessary to resolve the contradiction, as they represent a higher level assumption that belief in the elective is so necessary. It may not be possible to ensure this property for a particular elective because, as described in [Petrie 1987], the justification for an elective may produce an odd loop. In such a case, RAD will query the user as to whether the offending assertion may be safely omitted from the constructed justification. (This should be a relatively uncommon occurrence.) In addition, the elective is not permitted to receive the constructed justification if it would cause a previous contradiction to become IN.

The metarelation Support may be used to construct a justification for the elective. The elective is supported if Support is the relation symbol of a provable assertion with a single argument which unifies with the elective. In this case, the proven assertion will be added to the knowledge base with a justification based on its proof, and will be included in the IN-list of a new justification constructed for the elective. The OUT-list of this justification will contain the contradiction. In an intuitive sense, Support is the inverse of Defeat. The latter is used to argue against reason for belief in a culprit. Support is used to argue for belief in an elective. In terms of default reasoning, Support provides a way to indicate additional computation that should be performed to test the validity of an assumption. Initially, belief in the culprit may have been justified by failure of some limited attempt to prove the elective. If this culprit is later implicated in a contradiction, Support can be used to generate additional proof attempts. If the elective receives a Support justification, its semantics are that the elective received new support in order to resolve a contradiction.

If an elective cannot be proven and a new justification cannot be constructed for it, the CRM will attempt to validate one of the justifications of the elective, if any, by recursively making all of the data in the OUT-list OUT and in the IN-list IN. When trying to invalidate a justification and failing to make some elective from the OUT-list IN, the CRM will attempt to add elements to the OUT-list by proving any applicable *Defeat* assertions. The added elements will also become candidate electives. If no elective can be made IN, the CRM will pick some element of the IN-list as a new culprit to make OUT. No element will be chosen that is preferred to another.

As an example of this operation, suppose that, instead of the preferences above, the following query schema were defined for the example (see Chapter 9 for a description of ask):

```
ask((result(?x low ?y ?z)
    "Is the ~a of ~a for ~a low?"
        (?y ?z ?x) () t ())
    (mistaken-observation(?x ?y)
        "Were you mistaken in observing that ~a seemed ~a ?"
        (?x ?y) () t ())
    (test-error ?x ?y ?z ?u)
        "Was the test showing ~a to have a ~a ~a of ~a in error?"
        (?x ?y ?z ?u) () t ())
)
```

Then when Jane's lab-test results caused a contradiction, given the initial simple preference of belief in mistaken observations over that of lab tests, the CRM would first apply the Fix about the former, discover that there was a Defeat which made it applicable, and then try to prove the mistaken observation. There being no other way for the proof to succeed, and a query schemata having been defined for the assertion, the user would be queried

Were you mistaken in observing that JANE seemed DEHYDRATED?

If the user answered "no", the next Fix would be applied resulting in the query

Was the test showing JANE to have a LOW CONC of NA) in error?

A "no" answer to this query would exhaust the possible Fixes and cause dependency-directed backtracking to be attempted.

The supporters of the contradiction in this instance are the result of the lab test and the conclusion about high sodium. The OUT-list being empty, the CRM would attempt to make one of these assertions OUT. Since we already know that the lab test was not in error, the CRM would not be able to defeat the result of the lab test.

However, the conclusion of a high concentration of sodium rests upon an assumption of a normal amount of sodium. This assumption would be retraction if it could be believed that there was a low amount of sodium. Although there are no backward chaining rules with which to prove a low amount of sodium, there is a query schemata defined for this assertion. So the user would be queried:

#### Is the AMT of NA for JANE low?

An answer of either "yes" or "maybe" would resolve the contradiction. The "maybe" answer would produce the abductive justification that the preference above would. An answer of "no" would result in an unresolvable contradiction since there are no alternative electives except for the possibility of a mistaken observation, which has been eliminated by a previous query.

### 8.3 User Hints

We now give some useful hints for the use of contradiction resolution.

#### 8.3.1 General Assertions

If the top-level command wrong is used after a query about the inherited value of a multiple-valued slot for a frame, contradiction resolution will attempt to invalidate the justifications for the class slot value. This is also true for general assertions using relations. The semantics are that belief in the particular value for the slot rests on an assertion, the class slot value, that all instances of the class have that value for the slot. Contradiction of that value for one instance of that class entails contradiction of the more universal assertion that all instances have that value.

This is almost always a stronger contradiction than intended. In RAD, general assertions may not be defeated. As an example, suppose that we have a class of BANKER with a class value slot of wears (?X:banker watch) where Wears is a multiple-valued slot. If we want to contradict that a particular banker wears a watch, and the class value slot had a premise justification, resolution of the contradiction might be attempted with a Defeat such as:

Of course, there must be a way to prove such a defeat-elective. One way would be to provide a rule such as:

Such a rule would attempt to prove that each banker was timeless (rules or a query schemata must be given to allow such a proof). If it were able to do so for one, then all bankers would lose their watches. RAD does not permit this: the Defeat will essentially ignored.

If the user anticipates that such a class value slot for a multiple-valued slot might not hold for all instances of the class, but only represents a typical value, it is better to forego inheritance and make the slot values individual assumptions with a rule such as:

```
If (instance*(banker ?X)
    unless(timeless(?X))
then
   wears(?X watch)
```

This rule would allow individual bankers to lack watches without affecting others.

### 8.3.2 Retractable Default Assumptions

Default assumptions are those that rest on the lack of belief in assertions that are neither sufficiently important nor likely to make all attempts to find

a reason to believe them without further cause. The last rule of the previous section is intended to implement the default assumption that bankers typically wear watches. If the assertion that a banker is "timeless" is provable by use of a query schemata, then the user will be queried as to whether each banker is timeless. This is not what we desire in a default assumption. Rather we want to assume that a banker is not timeless unless we have specific reason to believe so. But if insufficient domain knowledge is supplied to justify the elective (typically an *Unless* condition) during contradiction resolution, then the assumption is not retractable.

There are three methods in RAD for deferring proof attempts on an elective in order to retract an assumption. The simplest is to use Known together with Unless to avoid any attempt to prove the Unless condition except for checking for a ground instance of the condition already in the database. The user may supply backward chaining rules which infer the condition, but they will not be invoked until contradiction resolution time.

The user may want to make limited proof attempts on the *Unless* condition. If so, *Known* is not used and the initial proof attempts can be represented by ordinary *RAD* backward chaining rules concluding the *Unless* condition. Further proof attempts which should be made only at contradiction resolution time can be made by writing *RAD* backward chaining rules which conclude the *Support* metarelation on the *Unless* condition. For example, if the assertion

## support(timeless(?X))

were provable, then the CRM would be able to defeat the assertion that some particular banker wears a watch. This *Support* may itself have been simply asserted, proven by a user query, or proven by more complex inferencing.

The other method of deferring proof attempts on an *Unless* condition is to omit the *Unless* clause and defer introduction of an elective in the knowledge base with *Defeats*. In this case, assumptions are created when needed by modification of OUT-lists. However, this only occurs during contradiction resolution. Prior to a *Defeat* being used, the assertion of the defeat-elective will not cause the defeat-culprit to go OUT since it has not yet been made an assumption resting on lack of belief in that particular elective. Thus *Defeats* should be used for rare defaults not usually considered: this introduces an element of heuristic inconsistency into the database.

86

Finally, electives need neither be proven nor *Support*ed but only assumed if it is provable that they are preferable to a contradiction. For example, if the assertion

```
Prefer(timeless(?X) contradiction)
```

were provable, the forward rule would also represent a retractable assumption. This uses of *Prefer* permits the CRM to generate an *abductive* justification as described in [Petrie 1987]. The general idea is that *reductio ad absurdem* reasoning is used. That a banker is timeless will be believed because it will cause a contradiction not to believe it.

#### 8.3.3 Preferences

Preferences are very general. Even though they are non-numeric, this does not mean that they may not be based on numbers. For instance, the following rule decides between two assertions based on their associated costs:

```
Infer Prefer(auto(?X) auto(?Y))
from (cost(?X ?Cost1)
      cost(?Y ?Cost2)
      ,<(?Cost1 ?Cost2))</pre>
```

As another example of generality, it is easy to express the knowledge that any element of some one set is preferable to that of another. Suppose that bigger(set-1 set-2) establishes a relationship between two sets. Then we may prefer elements of one set to another, based on that relationship with the rule:<sup>3</sup>

```
Infer Prefer(?X ?Y)
from (element(?X ?U)
    bigger(?U ?V)
    element(?Y ?V))
```

It is also easy to prefer one kind of assertion over any other. Suppose that assertions with relation *Boy* were to be preferred over any other. This would then be accomplished by asserting:

<sup>&</sup>lt;sup>3</sup>Warning: Transitivity is not guaranteed by such rules.

```
Prefer(boy(?X) ?Y)
```

When used to determine a choice between Fixes, it is important to remember that the initial choice of which Fix to pursue is based on the applicable fixelectives: as they have been instantiated by unification with the target. For instance, suppose we have:

```
Infer Prefer(lies(?X) jokes(?X))
from salesman(?X)
Fix(speaks(?X) known(?Y) jokes(?Z))
Fix(speaks(?X) known(?Y) lies(?Z))
```

If these Fixes were used to resolve a contradiction involving, say, the assertion speaks (Polly), the preference could not be used to decide between the two Fixes. the value of ?Z would be unbound at the time the choice of Fixes was to be made. This can be solved instead by concluding the preference which instantiates the possible liar or joker:

```
Infer Prefer(lies(?X) jokes(?X))
from (says(?X ?Y)
    unless(clown(?X))
    salesman(?X))
```

Finally, when trying to invalidate a justification of a culprit there is no way to express a preference for pursuing a particular element of the IN-list rather than one of the OUT-list. All elements of an OUT-list are chosen as electives and must fail before elements of the IN-list are examined. Preferences are only applicable in choosing between elements of an OUT-list or between elements of an IN-list. If there is another choice to be made, the user should restructure the knowledge representation.

## 8.4 Final Caveats

During contradiction resolution, the user may answer "yes-no" queries with "maybe". This has exactly the same effect as a "no" answer unless the query is about the belief in the elective that the CRM is currently trying to

prove. In that event, the elective receives a valid justification constructed by the CRM. This does not happen when the user answers "maybe" to a query about a goal generated by the attempted proof of the elective. In any case, if the elective cannot be proved and there is no query schemata for it, contradiction resolution will fail to use it to resolve the contradiction.

In general, the knowledge-guided contradiction resolution described here is a novel and complex method of computing. For the end-users of expert systems, it should provide additional functionality. The ability to disagree with a computer seems highly desirable. However, the novelty, complexity, and added functionality of this experimental computation can be expected to make the job of the knowledge engineer even more difficult. The interested RAD user is encouraged to try examples from this report and to experiment extensively before committing to development of an application. In any case, simple applications of the CRM, avoiding the full complexity and power of the system, are feasible.

# Chapter 9

# Syntax and the User Interface

The user interface is an interpreter that allows the user to enter commands, which are then executed by RAD. These commands allow the user to create, use, and modify knowledge bases. Commands can also be stored in files, then loaded into RAD and executed via a **consult** command.

In order to interact with agents, there must first be an Extensible Services Switch (ESS) running that provides communication and tree-space facilities. The ESS can be started by executing the following command on some workstation (assumed to be named "Delphi" here)

#### % ess -comm -client -tss

An arbitrary number of *RAD* agents and user interfaces can then be started. For an ESS on a host machine named "Delphi," an agent named "Zeus," and a user logged in as "User," this is done in the following two steps:

1. Create an agent by executing the following command in a window on "Delphi" or on another workstation

Connecting to ESS on Delphi+17001
Agent ZEUS is registered with the ESS.

RAD Agent Zeus is ready.

2. Create a user interface process in another window, or on another workstation, by

Connecting to ESS on Delphi+17001... ESS: Aide user-aide is ready.

User:

The prompt, User:, indicates who has said what or whose turn it is to talk.

The complete syntax for the user interface command is

```
raduser [-a agent]
     [-d]
     [-e ess-host [ess-port]]
     [-u user-name]
```

The options are

- -a initialize connection with the given agent.
- -d execute in debug mode; this mode is useful to yacc hackers.
- -e find the ess on the specified host and, optionally, at the port where it is known to be listening. For example, the command

% raduser -e Cronos 17002

will cause a search for the ess on the host Cronos at port 17002. Since we anticipate that users might often use the same ess, we have added a shell variable RADSERVER that is the name of the default ess host. So for example, if you put

#### setenv RADSERVER delphi

in your .cshrc, then you don't need to use the -e option unless you wish to override your default. Please notice that the host name must be in lower case.

-u use the given user-name for a prompt, instead of the default. The default is the value of the shell variable RADUSER, if present; otherwise, it is the value of the shell variable USER.

You can connect the user interface to a different ESS by using the ess command. In either case, once connected to an ESS, the user interface can directly communicate with any other agent or interface that is also connected to this ESS. You specify the particular agent with which you want to communicate by using the contact command. To communicate with an agent connected to an ESS different than your own, you must use the location command to tell your ESS where the agent is. This enables you to communicate with an agent executing anywhere reachable through OSI. The argument to the contact command is either an agent name or a two-element list whose first element is a host name and whose second is a port number. An example of how to use these commands is the following:

(assume that User is currently served by an ESS on delphi, and that agent Zeus is registered with the ESS on delphi.)
User: ess maui

User is unregistered from the ESS on delphi. Interface: Looking for ESS at maui port 17001. Connecting to ESS at maui+17001...

ESS: I am at maui port 17001. ESS: Aide user-aide is ready. User: location zeus delphi

User: list-agents

ESS: Registered agents and databases:

user is at maui zeus is at delphi

User: contact zeus

Interface: pinging agent zeus...

zeus: READY

User:

The ess command can also accept either a host name or a host-port pair.

If you are running raduser under a unix shell, then control-C will interrupt and restart it. If running under GNU Emacs, do control-C control-C instead.

## $9.1 \quad RAD \text{ Syntax}$

This section describes the syntax for RAD. A RAD program is a sequence of statements, each of which takes one of the following forms:

- A command, followed by its arguments. Some commands take a variable number of arguments; if there is more than one argument, the arguments must be enclosed in parentheses.
- A clause, such as an assertion or rule. In this case, an accept command is implicit.

Unit clauses take the form relation(args). Lists and conses are written as in Lisp. All escapes to external functions (written in Lisp or (eventually) C) are prefaced by a comma, but are otherwise written as unit clauses. (They are written this way, rather than in the syntax of their parent language, for the sake of compatibility should the externals be implemented differently.) Variables are prefaced by a question mark.

Backward rules have the form

Infer consequent from antecedents

while forward rules have the form

If antecedents then consequents

where consequent is a unit clause and antecedents and consequents are either unit clauses or lists thereof. Forward rules may have an optional priority associated with them, which, if present, appears in brackets before the If. Unit clauses and rules may also be given labels by which they may be referenced later. This is done by writing the label name, followed by a colon, before the clause

Users have great freedom in formatting their input as they like for the sake of readability. Input to RAD is not case-sensitive, except that case is preserved within literal strings, and for external tokens in the event that the external language is case-sensitive. Symbols can be written with any of the characters

Spaces, tabs, and newlines are mostly insignificant, except as follows:

- These characters normally serve to delimit lexical tokens. They can generally be omitted in cases where other punctuation is involved. For example, a rule label like "foo:" will be parsed as two tokens.
- Comments, which are prefaced by a semicolon, extend to the end of the input line.
- There can be no whitespace between a relation or function name and the left parenthesis that begins its argument list. This is necessary to distinguish between strings like "foo(bar)", which is a single term, and "foo (bar)" which is two. Likewise, there can be no whitespace between the name of a typed logical variable and its type specifier: "?x:baz" is a typed logical variable, whereas "?x:baz" is an untyped variable followed by a Lisp keyword.
- When a syntax error is detected, the parser applies a very simple heuristic: the remainder of the input line is discarded, and an attempt is made to begin parsing the next statement at the next input line.

The following example is a simple consult file that uses most of the features of the above syntax. Almost all operations, particularly changes to the class lattice, are in the form of assertions on appropriate primitive relations. Thus, a RAD consult file can consist mostly of clauses, with the accept being implicit.

```
;;; Echo assertions
echo t
;;; Turn syntax-checking off
check nil
;;; Consult two files
consult ("my-utilities.rad" "my-classes.rad")
;;; Create three new classes
assert-instance (class mammal)
assert-instance (class dog cat)
;;; Manipulate the class lattice
assert-subclass (mammal dog)
assert-subclass (mammal cat)
;;; Define append using backward rules
append(nil ?x ?x)
infer append((?x . ?tail) ?y (?x . ?newtail))
      from append(?tail ?y ?newtail)
;;; Create a forward rule with priority of 20
frule1: [20] if (a(?x) b(?x) c(?x))
             then d(?x)
;;; Reference an external function
foo-rule: infer foo(?x ?y ?z ?w)
          from = ((?x . ?y)
                 ,assoc(?z ?w :test #'eq)) ; Lisp escape
sys-rule: infer system-call(?x ?y)
          from =(?y ,system(?x)) ; C escape
```

### 9.2 The RAD Reader

The RAD reader is a collection of functions in the RAD agent that takes input from the parser and translates it into a form that is usable to the rest of the system. Part of the RAD reader's job is to dereference the names of objects that are passed to it. For example, in the following command, the strings "man" and "mortal" are taken to be the names of the relations Man and Mortal, respectively; these are defined if they do not already exist. This is because the RAD parser will recognize them as relations and will pass this information in the code that it generates for the agent.

User: Infer mortal(?x) from man(?x)

User: man(Socrates)

Likewise, the string "?x" is taken to be the name of a logical variable, ?X. As for "Socrates", it depends on whether or not "Socrates" is the name of something existing in the knowledge base. If SOCRATES were, for example, an instance of the class PHILOSOPHER, then the string would be taken to refer to that instance. However, if "Socrates" were not the name of any knowledge base object, it would simply be taken to stand for the symbol SOCRATES.

A common mistake, therefore, would be to define some string that has previously been interpreted as a symbol name to be the name of something else. Suppose, for example, the following were entered:

User: assert-instance (class philosopher)

User: man(socrates)

User: assert-instance (philosopher socrates)

User: ? man(socrates)

The query will unexpectedly fail. This is because the "Socrates" to which the query refers is the instance of **PHILOSOPHER**, whereas the one to which the assertion refers is just a symbol. (Of course, the query will work if the assertion and the assert-instance commands are interchanged.) RAD will

generate a warning if the user attempts to create an object with the same name as a symbol that it has previously encountered, but the operation will be performed anyway.

## 9.2.1 The Slash -/

The slash is used extensively in the names of a number of objects that are internal to RAD. Its use in the names of user-defined objects is therefore potentially dangerous and strongly discouraged.

#### 9.2.2 RAD Reader Macros

A number of syntactic constructs are implemented at the reader level by macros. These signal the reader to perform some kind of expansion on the input.

#### Question Mark -?

By itself, the question mark expands into the name of the function PROVE, which is an internal RAD function that implements a user query. Before another question mark, it expands into PROVE-ALL-INSTANCES, the RAD function that proves all instances of a goal.

A question mark appearing before anything else is, of course, taken to be the first character in the name of a logical variable. Avoid any use of the question mark that is not consistent with these conventions.

#### Comma -,

The comma is intended to provide a more concise and usable interface to Lisp than is provided by function declarations and the relations Is and Do. In general, the appearance of a comma in the antecedent of a forward or backward rule indicates that the following expression should be evaluated in Lisp. The reader examines the expression and generates appropriate calls to the relation Is. For example:

is translated internally to

```
Backward Rule Q-1 (IN)
Infer Q(?X ?Y) from ( IS(NIL (NOT (F ?X)))
IS(NIL (NOT (EVAL ?Y))) )
```

Commas can appear directly before a Lisp expression in the antecedent of a forward or backward rule, as in

```
Infer p(?x) from ,f(?x)
```

A proof of p(?x) will succeed as long as the lisp function (F?X) does not evaluate to NIL for the current binding of?X. Symbols can also be evaluated using the comma, although special forms cannot.

Expressions with commas can be used within relations, for example:

```
Infer p(?x) from q(,*foo*)
is translated to
Backward Rule P-2 (IN)
Infer P(?X) from ( IS(?/NEWO (EVAL (QUOTE *FOO*)))
```

Q(?/NEWO))

Note that the system automatically generates a logical variable to hold the result of the symbol evaluation.

Most importantly, however, the bindings of logical variables can be evaluated in Lisp. For instance,

```
Infer p(?x) from ,?x
Backward Rule P-3 (IN)
Infer P(?X) from IS(NIL (NOT (EVAL ?X)))
```

succeeds if the current binding of ?X is (CAR '(A)), or any Lisp expression that does not evaluate to nil.

Logical variables can appear anywhere in the scope of a comma. For instance, the following are all legal:

```
Infer q(?x ?y) from p( ,?x ,?y)
Infer p(?x) from ,f(?x)
Infer q(?x ?y) from p( ,f(?x) ,?y)
```

Functions that are referenced within the scope of a comma must be visible in the package PROT. This is true of symbols as well. The usual Common Lisp syntax conventions apply when referring to objects in other packages. Passing an unbound logical variable to Lisp, however, is not in itself an error. In this case, the variable will evaluate to itself.

Since commas get compiled into calls to Is, the user should observe the same care as if the use of Is were explicit. In the following example,

RULE-1 will never fire, since Is is not triggerable. Another problem can arise if all references to a logical variable in the head of a forward rule appear within the scope of a comma. Such a variable will never be bound to anything. For instance,

There is no way to assign a binding to ?X at top level. Hence the rule will not behave as one might expect.

The RAD reader allows only the first comma preceding a Lisp expression or logical variable to be a macro character. Subsequent commas revert to their traditional meaning, and must occur within backquoted expressions. For example,

```
Infer foo() from ,,X
```

produces an error. If a sequence of evaluations is desired, then it must be made explicit, to wit:

```
Infer foo() from ,(eval x)
```

Lisp macros can also be evaluated, but caution should be exercised since macro expansion takes place at rule compilation time rather than at proof time, even though this is not apparent from looking at the rule. For example, if the following macro is defined in lisp,

```
(defmacro bar (x) '(prog1 t (format t ''&Hi A'', x)))
```

then RAD would have the following behavior:

User: Infer baz(?x) from ,bar(?x)

Zeus: Backward Rule BAZ-1 (IN)

Infer BAZ(?X) from IS(NIL (NOT (BAR ?X)))

User: ? baz(mom)

Zeus: Hi MOM

Zeus: Yes.

If the macro is then redefined as follows

(defmacro bar (x) '(prog1 t (format t "~&Bye ~A" ,x)))

then RAD would have the following behavior:

User: ? baz(mom)

Zeus: Hi MOM

Zeus: Yes.

In order to produce the output "Bye MOM" one would have to erase BAZ-1 and reassert the same rule.

#### Unless

Although users can normally think of *Unless* as a built-in relation, it is implemented as a macro. Unfortunately, this is not completely transparent at the user interface level. For example,

Infer inanimate(?x) from unless(living(?x))

Backward Rule INANIMATE-1 (IN)

Infer INANIMATE(?X)

from

/UNLESS/0(?X PI::\*ARG-END-SYM\* LIVING)

/Unless/0 is a relation that is automatically created by macro expansion. Although this expansion is not transparent, users need not concern themselves as the semantics remain the same as if *Unless* were actually defined as a relation.

#### 9.2.3 Error Detection

In addition to the syntactic checking performed by the parser, the RAD reader tries to detect errors that result from input that is meaningless in light of the contents of the knowledge base. While some errors will still throw the user into the debugger, many common ones can be easily detected at the reader level.

Error detection is enabled at the start of radagent. Users can turn error detection off and on via the check command.

Errors fall into two classes, depending on their severity:

Warnings are printed when the user tries to do something questionable, where the action is otherwise syntactically correct and where aborting the action would be difficult or costly. The action is performed anyway. An example would be attempting to define a frame whose name had previously been used as an uninterpreted symbol.

Errors are printed when faulty syntax is detected before a command is processed. The command is not executed and control returns to RAD top level. An example would be supplying too many arguments to a top-level command.

## Warnings

• Attempt to redefine assumed symbol: A warning is printed if an object is created that shadows an uninterpreted symbol that the RAD reader has previously encountered. For example:

foo(bar)

User: assert-instance (class bar)

Zeus: Warning: BAR, previously assumed to be of type

## SYMBOL, is being redefined to be of type CLASS.

• Attempt to redefine object: RAD is creating an object that shadows an object previously defined by the user. For example:

User: f-1()

User: f()

Zeus: Syntax Warning: Attempt to redeclare or redefine the predicate F-1 to be of type ASSERTION.

The new definition will shadow the old.

Note that an actual error results if the user attempts to redefine an object (other than a symbol). See below.

• Relation arity violation: A relation has been supplied with the wrong number of arguments. This is normally an error (see below); a warning is generated only in the case of the ask and translate commands:

```
argument-types(baz (symbol))
```

translate baz(?x ?y) "~A is a ~A BAZ!" (?x ?y)

Syntax Warning: The arguments to the predicate BAZ, ?X ?Y, are in excess of its arity (1).

#### Errors

• Type error: An argument to a relation was of the wrong type.

User: assert-instance (relation foo)

User: argument-types(foo (class))

User: foo(5)

Zeus: The first argument to FOO, 5, is not of type CLASS.

• Attempt to redefine: The user has attempted to redeclare or redefine an object.

User: dog(5)

User: if d1() then assert-instance(relation dog-1)

User: d1()

Zeus: Attempt to declare Assertion DOG-1 to be a RELATION.

• Relation arity violation: A relation was referenced with the wrong number of arguments.

User: argument-types(foo (symbol))

User: foo(dog cat)

Zeus: Syntax Error: The arguments to the relation FOO,

DOG CAT, are in excess of its arity (1).

• Unrecognized type: The user qualified a logical variable to be an unknown type.

User: assert-instance (relation foo)

User: ? foo(?x:dawg)

Zeus: Syntax Error: DAWG is not a recognized type.

## 9.3 Top Level Commands

This section describes the top level commands available to the user within the stream interface. They can be entered from a keyboard or consulted by RAD from a file.

## 9.3.1 Creating and Modifying a Knowledge Base

accept pattern — (Erest patterns)

This command adds its arguments to the knowledge base, each with a justification consisting of the premise that an agent said the pattern and the

default that the agent is reliable, i.e., the agent cannot be proven to be unreliable. Note that the command name accept is typically omitted, without causing any ambiguity.

User: accept has-son(King Prince)

User: has-son(King Duke)

If the relation *Has-son* did not already exist in the knowledge base, it would be created and given an arity of two. However, if KING, Duke, and PRINCE did not already exist, they would be accepted as uninterpreted symbols by *RAD*.

An accept may also be a backward rule:

User: Mortality-rule: Infer mortal(?x) from man(?X)

Zeus: Backward Rule MORTALITY-RULE accepted.

Again, the relations *Man* and *Mortal* are created if they do not already exist. The user-supplied name, "Mortality-rule:", is optional.

Finally, an accept may be a forward rule:

User: If mother(?x ?y) then parent(?x ?y)

The rule may be preceded by a symbol to be used as the name of the rule and/or by a two-digit priority:

User: Parent-rule: [25] If mother(?x ?y) then parent(?x ?y)

assert pattern — (Erest patterns)

This command adds its arguments to the knowledge base, each with a premise justification. Note that the command name assert is typically omitted, without causing any ambiguity.

User: assert has-son(King Jester)
Zeus: Assertion HAS-SON-7 (IN)

An assertion may also be a backward or forward rule.

## assert-instance (class &rest names)

This command creates instances of the given class. If names is omitted, a name for the instance is generated by the system.

User: assert-instance (class dog cat)

User: show dog

Zeus: DOG is an instance of CLASS

Subclass of: none Subclasses: none Instances: none

## assert-subclass (super &rest subs)

This command causes super to be made an immediate superclass of each of the subs. All arguments must be existing classes.

User: assert-subclass (thing animal vegetable mineral)

User: show thing

Zeus: THING is an instance of CLASS

Subclass of:

none

Subclasses:

ANIMAL, VEGETABLE, MINERAL

Instances:

none

#### clear

This command reinitializes RAD by removing data and rules that are not system-defined.

## consult file — (Erest files)

This loads the indicated consult files into RAD. The files must be written as strings.

consult "test1.rad"

## contradict datum-name

This command takes an existing datum as an argument and contradicts it. See also wrong, which can be used to contradict the last proof result.

erase datum-name — ( &rest data)

The given data are to be made OUT. The JTMS then updates the support status of other data accordingly.

User: erase mortal-1

Zeus: OK.

User: why mortal-1

Zeus: Assertion MORTAL-1 (OUT)

MORTAL (SOCRATES)

JUSTIFICATION:

in-list NIL out-list NIL

was erased by assertion ERASED-1

User: show erased-1

Zeus: Assertion ERASED-1 (IN)

ERASED (MORTAL-1)

kill thing — (Erest things)

This deletes objects from the knowledge base. The objects are actually removed and their names are bound to a special symbol, DELETED. Contrast this with erase, which leaves data around, but puts them OUT. Moreover, erase only makes sense for data, i.e., things that actually have a support status, whereas kill can be used on data, relations, and frames.

User: kill (foo bar-1)

#### remember

When typed after a successful query, remember creates a datum representing the result of that query and asserts it. This is primarily useful for saving the result of a long proof so that the backward chainer would not have to repeat the proof.

User: ? likes(joe ?x)
Zeus: LIKES(JOE CANDY)

User: remember

Zeus: Remembered as Assertion LIKES-2.

User: show likes-2

Zeus: Assertion LIKES-2 (IN)

LIKES (JOE CANDY)

## remove-subclass (super &rest subs)

This removes all parent-child links between *super* and each element of *subs*. All arguments must be classes, of course.

User: remove-subclass (thing animal)

reply agent "clause"

Reply causes clause to be sent to the agent agent. The clause can be a datum, rule, or value, but it must be in the form of a string.

Zeus: P(A ?X)?

User: reply zeus "P(A 5)"

#### tell agent clause

Tell causes clause to be asserted in the knowledge base of the agent agent, as described in Chapter 2. The justification for clause has an IN-list consisting of reliable (SOURCE-AGENT) and said (SOURCE-AGENT CLAUSE), and an empty OUT-list, where SOURCE-AGENT is the name of the agent that executed the relation Tell. Note that clause may be a rule as well as a unit clause.

## toggle datum-name — ( Erest data)

This command changes the support status of datum-name from IN to OUT or OUT to IN, as appropriate. Any other datum dependent on datum-name may change its support status also.

#### wrong

This command is used to contradict the last goal proven by? or next. See also contradict, which can be used to contradict an existing datum.

# 9.3.2 Querying and Examining Knowledge Bases and Databases

#### ? goal

This command attempts to prove a goal by backward chaining. If the query succeeds the solution is printed, with appropriate variable bindings if variables were present in the goal pattern. If the query fails, then NO SOLUTION is printed.

User: age(hilary 7)

Zeus: Assertion AGE-1 accepted.

User: ? age(hilary ?x)
Zeus: AGE(HILARY 7)

#### ?? goal

This is like?, except that it constructs all proofs of a goal pattern. All successfully proved instances of the goal are output immediately.

#### explain

Explain, when typed after a successful query, shows why the query succeeded. For example,

User: ? mortal(socrates)
Zeus: MORTAL(SOCRATES)

User: explain

Zeus: MORTAL(SOCRATES)

was derived from the following: Assertion MAN-1 (IN)

MAN(SOCRATES)

Backward Rule MORTAL-1 (IN)
Infer MORTAL(?X) from MAN(?X)

#### next

When typed after a successful call to ?, next initiates another proof attempt of the same goal in order to derive another solution, until all instances of the goal have been derived.

User: ? age(?x 7)
Zeus: AGE(HILARY 7)

User: next

Zeus: AGE(JENNIFER 7)

User: next

Zeus: NO SOLUTION

query agent goal

Query succeeds if goal is provable by the agent agent. The agent returns all possible proofs of goal, which are then used one at a time. For each successful proof, an entry for the proof is created in a table of remote proofs, with a pointer to agent who supplied the proof.

## query-once agent goal

Query-once succeeds if goal is provable by the agent agent. If so, it creates an entry for the proposition in its table of remote proofs, with a pointer to agent who supplied the proof.

#### db-query database pattern

Db-query succeeds if the database database contains any tuples that match pattern. The database returns all possible matching tuples, which are then used one at a time by the agent that issued the db-query command. Internally, the interface converts this command into an equivalent query in SQL syntax that is processed according to the RDA protocol. For this query to succeed, the database must already be opened (see the commands db-connect and db-disconnect).

User: contact zeus

Zeus: READY

User: db-connect TravelDB delphi

User: db-query TravelDB hotelinventory(?hotel ?city)

Zeus: hotelinventory("Hilton" "San Diego")

hotelinventory("Sheraton" "Denver")

hotelinventory("Hyatt" "Austin")

User: db-disconnect TravelDB

show object — (Erest objects)

Show displays the printed representation of its arguments, which may be data, relations, or frames.

User: show age-1

Zeus: Assertion AGE-1 (IN)

AGE(HILARY 7)

User: show dog

Zeus: DOG is an instance of CLASS

Subclass of: none Subclasses: none Instances: FIDO

110

#### why datum

This command explains the support status of a datum. If the datum is IN, the IN- and OUT-lists of its justification are printed. If it is OUT, but has justifications, then the data that invalidate its justifications are printed.

```
User: why mortal-1
                             (IN)
Zeus: Assertion MORTAL-1
  MORTAL (SOCRATES)
was derived from the following:
Forward Rule FRULE-0
  If
   MAN(?X)
  then
   MORTAL (?X)
Assertion MAN-1
                    (IN)
  MAN(SOCRATES)
User: erase mortal-1
Zeus: OK.
User: why mortal-1
Zeus: Assertion MORTAL-1
                             (TUO)
  MORTAL (SOCRATES)
JUSTIFICATION:
    in-list
                (FRULE-0 MAN-1)
    out-list
                NIL
was erased by Assertion ERASED-1
```

## 9.3.3 Customizing the Interface

ask pattern string bound-vars unbound-vars t-or-nil prompts

This command creates an object called a query schema, which determines a format in which the user may be prompted for information during the course of a proof, by means of the built-in relations Ask and Ask-once (see Chapter

6). A query schema consists of the following components:

pattern is a clause, possibly containing variables, with which the query schema is associated.

string is a string to be used as the second argument to the Lisp function FORMAT; it may contain occurrences of ~A.

bound-vars is a list of some of the variables that occur in pattern; its length must be the number of occurrences of "A in string.

unbound-vars is a list of the variables occurring in pattern that do not belong to bound-vars.

t-or-nil is either T or NIL, depending on whether or not the query determined by string requires a yes-or-no response.

prompts is a list of strings of the same length as unbound-vars.

A query schema is relevant to a goal if the goal is an instance of its pattern with each variable in bound-vars replaced by a constant. When the user is to be queried for an instance of a goal that is being processed by the backward chainer, the list of query schemata associated with the relation of the goal is examined until either the end of the list is reached or one is found that is relevant to the goal. In the first case, the default format (see the definition of Ask) is used for the query. Otherwise, a format is determined by the relevant query schema as described below.

As an example, consider the command

```
ask birthplace(?X ?Y ?Z) "Enter the birthplace of "A" (?X) (?Y ?Z) NIL (" city: " " state: ")
```

Suppose that after this command has been issued, the user is to be queried for an instance of the goal BIRTHPLACE(MICHAEL ?C ?S). Since this goal is an instance of the pattern BIRTHPLACE(?X ?Y ?Z) with ?X (the only variable in bound-vars) replaced by the constant MICHAEL, the query schema defined by the command is relevant. The string is then printed with ~A replaced by MICHAEL. The user is then prompted for values of ?Y and ?Z:

Enter the birthplace of MICHAEL

city: BOSTON

state: MASSACHUSETTS

In this case, the resulting assertion

BIRTHPLACE (MICHAEL BOSTON MASSACHUSETTS)

is successfully matched against the goal.

If t-or-nil is T, as in

ask birthplace(?X ?Y ?Z) "Do you know where ~A was born?"
(?X) (?Y ?Z) T (" city: " " state: ")

then a yes-or-no response is expected after *string* is printed, determining whether to proceed with the query:

Do you know where MICHAEL was born? (Y or N) Yes

city: BOSTON

state: MASSACHUSETTS

Do you know where BOB was born? (Y or N) No

Query schemata are matched against a goal in the order in which they were originally defined. If several query schemata are relevant to a goal, than this order is significant. Generally, they should be entered in order of decreasing specificity.

ask birthplace(?X ?Y ?Z) "Was ~A born in ~A,~A?" (?X ?Y ?Z) () T ()

ask birthplace(?X ?Y ?Z) "Do you know where ~A was born?" (?X) (?Y ?Z) T (" city: " state: ")

ask birthplace(?X ?Y ?Z) "Do you know anyone born in ~A?" (?Z) (?X ?Y) T (" who? " " what city? ")

In this case, all three schemata are relevant to the goal BIRTHPLACE(JIM CLEVELAND OHIO), but only the first would be used in querying about this goal.

#### check t-or-nil

The check command controls whether or not the agent checks for errors in the parser input. It takes one argument, which can be T or NIL: check nil suppresses the checking for and printing of errors and warnings, while check t turns error detection back on. Note that this applies only to the agent: the syntax checking performed by the parser cannot be disabled.

#### echo t-or-nil

This command turns on the echoing of user-defined assertions and rules. This behavior is enabled by default when RAD starts up, and can be shut off with echo nil.

#### translate pattern string vars

This command specifies translation schema to be associated with certain patterns, in order that they can be printed in a more readable form. Each schema is specified by three things:

- A pattern to be translated when printed.
- A string specifying the translation. This should be a string acceptable to the format function, with "A's used to mark the place of logical variables.
- A list of variables that are passed as arguments to format. These must be a subset of the variables appearing in pattern.

For example,

```
translate age(?x 1) ""A is a year old." (?x)
```

```
translate age(?x ?y) "~A is ~A years old." (?x ?y)
```

creates two query schema associated with the relation Age. If one were to assert AGE(GEORGE 15), the following would be printed:

```
Assertion AGE-1 (IN)
GEORGE is 15 years old.
```

Note that the translation schema will be matched in the order they are declared. Users should therefore enter the most specific schema first.

## 9.3.4 Miscellaneous Commands

#### bgrind reln

This takes a relation reln and outputs the WAM code generated for its backward rules.

#### fgrind reln

This command takes a relation reln and outputs the WAM code generated for its forward rules.

#### contact arg

The contact command is used to connect the user interface to different agents. For agents connected to a different ESS than the one to which you are connected, you must first issue the location command. This enables a user to communicate with an agent executing anywhere that is reachable by TCP/IP or OSI. The argument to the contact command is either an agent name or a two-element list whose first element is a host name and whose second is a port number.

## db-connect database host

The db-connect command is used to connect the user interface to a database. This causes an open-db instruction in the RDA protocol to be executed. The arguments to the db-connect command are a database name and the name of the host where the database is located.

#### db-disconnect database

The db-disconnect command is used to disconnect the user interface from a database. This causes an close-db instruction in the RDA protocol to be executed. The argument to the db-disconnect command is a database name.

#### ess arg

The ess command is used to connect the user interface to an ess and, thereby, directly to the set of agents that the ess serves. The argument to the ess command is either a host name or a two-element list whose first element is a host name and whose second is a port number.

#### location arg

The location command is used to connect the user interface to agents that are connected to and serviced by other ESS's. The two arguments to the location command are an agent name and the host where that agent's ESS is executing.

#### halt-agent agent

The halt-agent command is used to stop the execution of an agent with which you are in contact (see the contact command). The argument to this command is the name of the agent to be halted. The agent unregisters itself from its ESS and returns control to the operating system.

## list-agents

The list-agents command is used to display all of the agents and databases that are known to the ESS, for example

User: list-agents

ESS: Registered agents and databases:

user is at delphi zeus is at delphi traveldb is at olympus

User:

#### help command

This command provides a simple online help facility for most commands.

User: help assert-instance

Zeus: ASSERT-INSTANCE (Class & Rest Names)

## quit

This exits the user interface and returns the user to the operating system. This does not affect the operation of any agents with which the user interface may have been communicating.

# Chapter 10

## Installation

## 10.1 Environment

The Distributed Communicating Agent software currently operates on the following system architectures:

- Sun 4 with SunOS 4.x and Allegro 4.0.1 Common Lisp
- Sun 3 with SunOS 4.x and Allegro 4.0.1 Common Lisp
- DECstation 5100 with Ultrix V4.2 and Lucid Common Lisp Version 4.0

In order to build the Distributed Communicating Agent software, you must have the following software tools installed:

- C development tools, e.g., cc, libc.a, and include files
- C library containing socket system calls, if not already in libc.a
- Common Lisp

Additionally, in order to execute the software, you must have a Rosette Extensible Services Switch (ESS) executing with its client, communication, and tree-space options available.

## 10.2 Sources

The source files for the Distributed Communicating Agent software are in four directories. The files are described relative to a Carnot root directory, typically /usr/carnot. The shell ENVIRONMENT variable CARNOTDIR must be set to this prior to making any of the components, e.g.,

#### % setenv CARNOTDIR /usr/carnot

The four directories, depicted in Figure 10.1, contain

- 1. sources for RAD agents,
  - "carnot/semantic/dca/release/base/src/agent/\*
- 2. sources for RAD interfaces (human agents),
  - "carnot/semantic/dca/release/base/src/interface/\*
- 3. a library of definitions for built-in RAD commands and funtions,
  - ~carnot/semantic/dca/release/base/src/radlib/\*
- 4. a library of example knowledge bases, some of which are described in the appendix to this document
  - ~carnot/lib/dca/\*

## 10.3 The MAKE Process

Installation of the Distributed Communicating Agent software is automated by Makefiles. However, you need to define several MAKE variables for your particular environment. The carnot/config directory contains configuration Makefiles for the supported architectures. Examine the appropriate file and modify it for your environment. In particular, set the value of the variable RADLISP to the path where your lisp is located. Then run the carnot/Makefile. For example, to compile and install the software on a Sun4, do the following from CARNOTDIR:

% cmake semantic-svcs

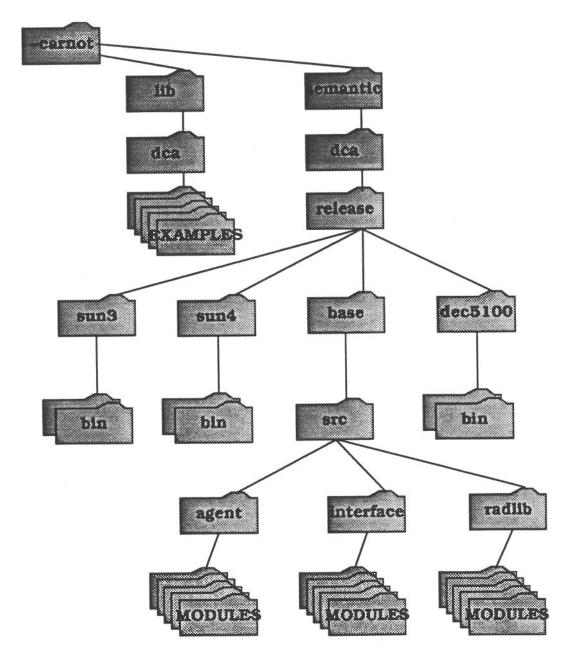


Figure 10.1: Organization of the Distributed Communicating Agent software

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# Appendix A

# Multiagent Tic-Tac-Toe

This section of the document describes a RAD implementation of tic-tactoe involving four agents: a referee, two players, and a human observer. The human observer communicates with the referee to initiate the game. The referee 1) determines if the two players are ready to play the game, 2) finds out if the players have a preference for "X" or "O" and assigns marks accordingly, 3) "flips a coin" to determine which player goes first, 4) successively asks each player for his next move, 5) checks each move for legality, 6) determines when a game is over, and 7) declares a winner (if any). The players each have knowledge about tic-tac-toe and a strategy for playing it.

## A.1 Knowledge Base for Referee

;;\*\*\* Class and Instance Declarations \*\*\*

assert-instance(class player)
assert-subclass(agent player)
assert-instance(player PlayerA PlayerB)
assert-instance(agent referee)
assert-instance(class game)
assert-instance(class nought-or-cross)
assert-instance(nought-or-cross 0 X)
assert-instance(class index)

```
assert-instance(index zero one two)
assert-instance(class agent-capability)
assert-instance(agent-capability tic-tac-toe-playing)
assert-instance(game game1)
;;*** Relation Declarations ***
assert-instance(relation mod1)
argument-types(mod1 (index index))
assert-instance(relation mod2)
argument-types(mod2 (index index))
assert-instance(relation expertise); multiple-valued
argument-types(expertise (agent agent-capability))
specificity-ordered(expertise)
assert-instance(relation awake)
argument-types(awake (agent symbol))
single-valued(awake)
specificity-ordered(awake)
assert-instance(relation abnormal)
argument-types(abnormal (agent))
assert-instance(relation play-tic-tac-toe)
argument-types(play-tic-tac-toe (player player game))
assert-instance(relation tic-tac-toe)
argument-types(tic-tac-toe (player player game))
assert-instance(relation mark)
argument-types(mark (player game nought-or-cross))
single-valued(mark)
specificity-ordered(mark)
assert-instance(relation prefer-mark)
```

```
argument-types(prefer-mark (player nought-or-cross))
single-valued(prefer-mark)
specificity-ordered(prefer-mark)
assert-instance(relation opposite-mark)
argument-types(opposite-mark (nought-or-cross nought-or-cross))
single-valued(opposite-mark)
specificity-ordered(opposite-mark)
assert-instance(relation coin-toss-result)
argument-types(coin-toss-result (game number))
single-valued(coin-toss-result)
specificity-ordered(coin-toss-result)
assert-instance(relation goes-next)
argument-types(goes-next (game fixnum player))
specificity-ordered(goes-next)
assert-instance(relation other-player)
argument-types(other-player (player player game))
single-valued(other-player)
specificity-ordered(other-player)
assert-instance(relation board)
argument-types(board (game index index nought-or-cross))
specificity-ordered(board)
assert-instance(relation game-over)
argument-types(game-over (game))
specificity-ordered(game-over)
assert-instance(relation tie-game)
argument-types(tie-game (game))
specificity-ordered(tie-game)
assert-instance(relation wins-game)
argument-types(wins-game (player game))
```

```
assert-instance(relation move)
argument-types(move (player index index game))
specificity-ordered(move)
assert-instance(relation okay-move)
        argument-types(okay-move (game player index index))
specificity-ordered(okay-move)
;;*** End of Relation Declarations ***
        mod1(zero one)
        mod1(one two)
        mod1(two zero)
        mod2(zero two)
        mod2(one zero)
        mod2(two one)
expertise(referee tic-tac-toe-playing)
opposite-mark(X 0)
opposite-mark(O X)
Infer awake(?Agent Yes)
from unless(abnormal(?Agent))
;; Begin a game of tic-tac-toe between two Players.
;; Make sure the Players are ready to play...
Begin-tic-tac-toe:
  If (play-tic-tac-toe(?Player1 ?Player2 ?Game)
       query-once( ?Player1 awake(?Player1 Yes))
       query-once( ?Player2 awake(?Player2 Yes)))
  then (tic-tac-toe(?Player1 ?Player2 ?Game)
        other-player(?Player1 ?Player2 ?Game)
        other-player(?Player2 ?Player1 ?Game)
```

specificity-ordered(wins-game)

```
tell( ?Player1 other-player(?Player1 ?Player2 ?Game))
       tell( ?Player2 other-player(?Player2 ?Player1 ?Game)))
Trigger-mark-choice:
 If (tic-tac-toe(?Player1 ?Player2 ?Game)
      excuse(unless( marks-chosen(?Game))))
 then (preference-rule-enabled(?Game)
                                        ;Order of consequents
       no-preference-rule-enabled(?Game)); determines order of
                                         ;rule firing.
Choose-marks-1: ;give Player1 his preference
  If (tic-tac-toe(?Player1 ?Player2 ?Game)
       excuse(unless(marks-chosen(?Game)))
      preference-rule-enabled(?Game)
      query-once( ?Player1 prefer-mark(?Player1 ?Mark))
       opposite-mark(?Mark ?Other-mark))
  then (mark(?Player1 ?Game ?Mark)
       mark(?Player2 ?Game ?Other-mark)
        tell( ?Player1 mark(?Player1 ?Game ?Mark))
        tell( ?Player1 mark(?Player2 ?Game ?Other-mark))
        tell( ?Player2 mark(?Player1 ?Game ?Mark))
        marks-chosen(?Game))
Choose-marks-2: ;give Player2 his preference
      (tic-tac-toe(?Player1 ?Player2 ?Game)
       excuse(unless( marks-chosen(?Game)))
       preference-rule-enabled(?Game)
       query-once( ?Player2 prefer-mark(?Player2 ?Mark))
       opposite-mark(?Mark ?Other-mark))
  then (mark(?Player2 ?Game ?Mark)
        mark(?Player1 ?Game ?Other-mark)
        tell( ?Player1 mark(?Player2 ?Game ?Mark))
        tell( ?Player1 mark(?Player1 ?Game ?Other-mark))
        tell( ?Player2 mark(?Player2 ?Game ?Mark))
        tell( ?Player2 mark(?Player1 ?Game ?Other-mark))
```

```
Choose-marks-3: ;assign marks if no preferences
 If (tic-tac-toe(?Player1 ?Player2 ?Game)
       excuse(unless( marks-chosen(?Game)))
       no-preference-rule-enabled(?Game))
  then (mark(?Player1 ?Game X)
        mark(?Player2 ?Game 0)
        tell( ?Player1 mark(?Player1 ?Game X))
        tell( ?Player1 mark(?Player2 ?Game 0))
        tell( ?Player2 mark(?Player1 ?Game X))
        tell( ?Player2 mark(?Player2 ?Game 0))
        marks-chosen(?Game))
Decide-who-goes-first: ;the referee "flips a coin"
  If (tic-tac-toe(?Player1 ?Player2 ?Game)
       marks-chosen(?Game)
       excuse(unless(coin-toss-result(?Game ?Number))))
  then coin-toss-result(?Game ,random(1.0)) ;0 <= ?X < 1.0
  If (tic-tac-toe(?Player1 ?Player2 ?Game)
       coin-toss-result(?Game ?X)
       ,>=(?X 0.5)) ;Player1 goes first
  then goes-next(?Game 1 ?Player1)
  If (tic-tac-toe(?Player1 ?Player2 ?Game)
       coin-toss-result(?Game ?X)
        .<(?X 0.5)) ;Player2 goes first</pre>
  then goes-next(?Game 1 ?Player2)
Incorporate-next-player-move:
  If (goes-next(?Game ?Move-number ?Player1); Next player moves
       excuse(unless(goes-next(?Game ,1+(?Move-number) ?Anyone)))
                                               :unless game over
       excuse(unless(game-over(?Game)))
       ,<(?Move-number 10)</pre>
       okay-move(?Game ?Player1 ?RMove ?CMove);Get next move
       mark(?Player1 ?Game ?Mark)
```

marks-chosen(?Game))

```
other-player(?Player1 ?Player2 ?Game))
 then (board(?Game ?RMove ?CMove ?Mark)
        tell( ?Player1 board(?Game ?RMove ?CMove ?Mark))
        tell( ?Player2 board(?Game ?RMove ?CMove ?Mark))
        print(""%Move "A: square "A "A is occupied by "A"
      (?Move-number ?RMove ?CMove ?Player1))
        goes-next(?Game ,1+(?Move-number) ?Player2))
                                     ;Other player is next
Find-good-move:
  Infer okay-move(?Game ?Player ?RMove ?CMove)
  from (query-once(?Player move(?Player ?RMove ?CMove ?Game))
        excuse(unless(board(?Game ?RMove ?CMove ?Any-mark))))
Test-row-1:
  If (mark(?Player ?Game ?Mark)
      board(?Game zero zero ?Mark)
      board(?Game zero one ?Mark)
      board(?Game zero two ?Mark))
  then (wins-game(?Player ?Game)
        game-over(?Game))
Test-row-2:
  If (mark(?Player ?Game ?Mark)
      board (?Game one zero ?Mark)
      board(?Game one one ?Mark)
      board(?Game one two ?Mark))
  then (wins-game(?Player ?Game)
        game-over(?Game))
Test-row-3:
  If (mark(?Player ?Game ?Mark)
      board(?Game two zero ?Mark)
      board(?Game two one ?Mark)
      board(?Game two two ?Mark))
  then (wins-game(?Player ?Game)
        game-over(?Game))
```

```
If (mark(?Player ?Game ?Mark)
      board(?Game zero zero ?Mark)
      board(?Game one zero ?Mark)
      board(?Game two zero ?Mark))
 then (wins-game(?Player ?Game)
        game-over(?Game))
Test-column-2:
  If (mark(?Player ?Game ?Mark)
      board(?Game zero one ?Mark)
      board(?Game one one ?Mark)
      board(?Game two one ?Mark))
  then (wins-game(?Player ?Game)
        game-over(?Game))
Test-column-3:
  If (mark(?Player ?Game ?Mark)
      board(?Game zero two ?Mark)
      board(?Game one two ?Mark)
      board(?Game two two ?Mark))
  then (wins-game(?Player ?Game)
        game-over(?Game))
Test-diagonal-1:
  If (mark(?Player ?Game ?Mark)
      board(?Game zero zero ?Mark)
      board(?Game one one ?Mark)
      board(?Game two two ?Mark))
  then (wins-game(?Player ?Game)
        game-over(?Game))
Test-diagonal-2:
  If (mark(?Player ?Game ?Mark)
      board(?Game zero two ?Mark)
      board (?Game one one ?Mark)
```

Test-column-1:

```
board(?Game two zero ?Mark))
 then (wins-game(?Player ?Game)
       game-over(?Game))
Test-tie-game:
  If (board(?Game zero zero ?Any-mark1)
      board(?Game zero one ?Any-mark2)
      board(?Game zero two ?Any-mark3)
      board(?Game one zero ?Any-mark4)
      board(?Game one one ?Any-mark5)
      board(?Game one two ?Any-mark6)
      board(?Game two zero ?Any-mark7)
      board(?Game two one ?Any-mark8)
      board(?Game two two ?Any-mark9))
  then (tie-game(?Game)
        game-over(?Game))
Print-message-tie:
  If tie-game(?Game)
  then (print("~2%Game ~A ends in a tie." (?Game)))
Print-message-win:
  If wins-game(?Player ?Game)
  then (print("~2%Game ~A is won by ~A." (?Game ?Player)))
Print-message-over:
  If game-over(?Game)
  then (print("~2%Game ~A is over." (?Game)))
        Knowledge Base for Player A
;;*** Class and Instance Declarations ***
assert-instance(class player)
assert-subclass(agent player)
assert-instance(player PlayerA PlayerB)
                                                 MCC Carnot Project
                                  129
```

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```
assert-instance(agent referee)
assert-instance(class game)
assert-instance(class nought-or-cross)
assert-instance(nought-or-cross 0 X)
assert-instance(class index)
assert-instance(index zero one two)
assert-instance(class agent-capability)
assert-instance(agent-capability tic-tac-toe-playing)
;;*** Relation Declarations ***
assert-instance(relation mod1)
argument-types(mod1 (index index))
assert-instance(relation mod2)
argument-types(mod2 (index index))
assert-instance(relation expertise)
argument-types(expertise (agent agent-capability))
specificity-ordered(expertise)
assert-instance(relation awake)
argument-types(awake (agent symbol))
single-valued(awake)
specificity-ordered(awake)
assert-instance(relation abnormal)
argument-types(abnormal (agent))
assert-instance(relation play-tic-tac-toe)
argument-types(play-tic-tac-toe (player player game))
assert-instance(relation tic-tac-toe)
argument-types(tic-tac-toe (player player game))
assert-instance(relation mark)
argument-types(mark (player game nought-or-cross))
                                                 MCC Carnot Project
                                  130
```

Confidential and Proprietary

```
single-valued(mark)
specificity-ordered(mark)
assert-instance(relation prefer-mark)
argument-types(prefer-mark (player nought-or-cross))
single-valued(prefer-mark)
specificity-ordered(prefer-mark)
assert-instance(relation opposite-mark)
argument-types(opposite-mark (nought-or-cross nought-or-cross))
single-valued(opposite-mark)
specificity-ordered(opposite-mark)
assert-instance(relation other-player)
argument-types(other-player (player player game))
single-valued(other-player)
specificity-ordered(other-player)
assert-instance(relation board)
argument-types(board (game index index nought-or-cross))
specificity-ordered(board)
assert-instance(relation game-over)
argument-types(game-over (game))
specificity-ordered(game-over)
assert-instance(relation tie-game)
argument-types(tie-game (game))
specificity-ordered(tie-game)
assert-instance(relation wins-game)
argument-types(wins-game (player game))
 specificity-ordered(wins-game)
 assert-instance(relation move)
 argument-types(move (player index index game))
 specificity-ordered(move)
```

```
;;*** End of Relation Declarations ***
        mod1(zero one)
        mod1(one two)
        mod1(two zero)
        mod2(zero two)
        mod2(one zero)
        mod2(two one)
expertise(referee tic-tac-toe-playing)
opposite-mark(X 0)
opposite-mark(0 X)
Infer awake(?Agent Yes)
from unless(abnormal(?Agent))
;;;;;;;;;;; Rules that Generate Suggested Moves ;;;;
Block-row:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row ?Col1 ?Mark)
        board(?Game ?Row ?Col2 ?Mark)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Block-column:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row1 ?Col ?Mark)
        board(?Game ?Row2 ?Col ?Mark)
        mod1(?Row ?Row1)
```

```
mod2(?Row ?Row2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Block-principal-diagonal:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row1 ?Col1 ?Mark)
        board(?Game ?Row2 ?Col2 ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Block-anti-principal-diagonal:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row1 ?Col2 ?Mark)
        board(?Game ?Row2 ?Col1 ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-row-next-own:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Opponent-mark)
        mark(?Player ?Game ?Mark)
        board(?Game ?Row ?Col1 ?Mark)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
         excuse(unless(board(?Game ?Row ?Col2 ?Opponent-mark)))
         excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
```

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```
Try-row-next-to-next-own:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Opponent-mark)
        mark(?Player ?Game ?Mark)
        board(?Game ?Row ?Col2 ?Mark)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col1 ?Opponent-mark)))
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-column-next-own:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Opponent-mark)
        mark(?Player ?Game ?Mark)
        board(?Game ?Row1 ?Col ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
         excuse(unless(board(?Game ?Row2 ?Col ?Opponent-mark)))
         excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-column-next-to-next-own:
   Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
         mark(?Opponent ?Game ?Opponent-mark)
         mark(?Player ?Game ?Mark)
         board(?Game ?Row2 ?Col ?Mark)
         mod1(?Row ?Row1)
         mod2(?Row ?Row2)
         excuse(unless(board(?Game ?Row1 ?Col ?Opponent-mark)))
         excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
 Select-center:
   Infer move(?Player one one ?Game)
   from (excuse(unless(board(?Game one one ?Any-mark))))
```

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```
Select-corner-00:
  Infer move(?Player zero zero ?Game)
  from (excuse(unless(board(?Game zero zero ?Any-mark))))
Select-corner-02:
  Infer move(?Player zero two ?Game)
  from (excuse(unless(board(?Game zero two ?Any-mark))))
Select-corner-20:
  Infer move(?Player two zero ?Game)
  from (excuse(unless(board(?Game two zero ?Any-mark))))
Select-corner-22:
  Infer move(?Player two two ?Game)
  from (excuse(unless(board(?Game two two ?Any-mark))))
Select-middle-01:
  Infer move(?Player zero one ?Game)
  from (excuse(unless(board(?Game zero one ?Any-mark))))
Select-middle-10:
  Infer move(?Player one zero ?Game)
  from (excuse(unless(board(?Game one zero ?Any-mark))))
Select-middle-12:
  Infer move(?Player one two ?Game)
  from (excuse(unless(board(?Game one two ?Any-mark))))
Select-middle-21:
  Infer move(?Player two one ?Game)
  from (excuse(unless(board(?Game two one ?Any-mark))))
assert-instance(game game1)
prefer-mark(PlayerA X)
```

#### A.3 Knowledge Base for Player B

```
;;*** Class and Instance Declarations ***
assert-instance(class player)
assert-subclass(agent player)
assert-instance(player PlayerA PlayerB)
assert-instance(agent referee)
assert-instance(class game)
assert-instance(class nought-or-cross)
assert-instance(nought-or-cross 0 X)
assert-instance(class index)
assert-instance(index zero one two)
assert-instance(class agent-capability)
assert-instance(agent-capability tic-tac-toe-playing)
;;*** Relation Declarations ***
assert-instance(relation mod1)
argument-types(mod1 (index index))
assert-instance(relation mod2)
argument-types(mod2 (index index))
assert-instance(relation expertise)
argument-types(expertise (agent agent-capability))
specificity-ordered(expertise)
assert-instance(relation awake)
argument-types(awake (agent symbol))
single-valued(awake)
specificity-ordered(awake)
assert-instance(relation abnormal)
argument-types(abnormal (agent))
assert-instance(relation play-tic-tac-toe)
```

```
argument-types(play-tic-tac-toe (player player game))
assert-instance(relation tic-tac-toe)
argument-types(tic-tac-toe (player player game))
assert-instance(relation mark)
argument-types(mark (player game nought-or-cross))
single-valued(mark)
specificity-ordered(mark)
assert-instance(relation prefer-mark)
argument-types(prefer-mark (player nought-or-cross))
single-valued(prefer-mark)
specificity-ordered(prefer-mark)
assert-instance(relation opposite-mark)
argument-types(opposite-mark (nought-or-cross nought-or-cross))
single-valued(opposite-mark)
specificity-ordered(opposite-mark)
assert-instance(relation other-player)
argument-types(other-player (player player game))
single-valued(other-player)
specificity-ordered(other-player)
assert-instance(relation board)
argument-types(board (game index index nought-or-cross))
specificity-ordered(board)
assert-instance(relation game-over)
argument-types(game-over (game))
specificity-ordered(game-over)
assert-instance(relation tie-game)
argument-types(tie-game (game))
specificity-ordered(tie-game)
```

```
assert-instance(relation wins-game)
argument-types(wins-game (player game))
specificity-ordered(wins-game)
assert-instance(relation move)
argument-types(move (player index index game))
specificity-ordered(move)
;;*** End of Relation Declarations ***
        mod1(zero one)
        mod1(one two)
        mod1(two zero)
        mod2(zero two)
        mod2(one zero)
        mod2(two one)
expertise(referee tic-tac-toe-playing)
opposite-mark(X 0)
opposite-mark(0 X)
Infer awake(?Agent Yes)
from unless(abnormal(?Agent))
;;;;;;;;;;; Rules that Generate Suggested Moves ;;;;
Block-row:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row ?Col1 ?Mark)
        board(?Game ?Row ?Col2 ?Mark)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
```

```
Infer move(?Player ?Row ?Col ?Game)
 from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row1 ?Col ?Mark)
        board(?Game ?Row2 ?Col ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Block-principal-diagonal:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row1 ?Col1 ?Mark)
        board(?Game ?Row2 ?Col2 ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Block-anti-principal-diagonal:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Mark)
        board(?Game ?Row1 ?Col2 ?Mark)
        board(?Game ?Row2 ?Col1 ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-row-next-own:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
```

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Block-column:

```
mark(?Opponent ?Game ?Opponent-mark)
       mark(?Player ?Game ?Mark)
        board(?Game ?Row ?Col1 ?Mark)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col2 ?Opponent-mark)))
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-row-next-to-next-own:
  Infer move(?Player ?Row ?Col ?Game)
 from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Opponent-mark)
        mark(?Player ?Game ?Mark)
        board(?Game ?Row ?Col2 ?Mark)
        mod1(?Col ?Col1)
        mod2(?Col ?Col2)
        excuse(unless(board(?Game ?Row ?Col1 ?Opponent-mark)))
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-column-next-own:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Opponent-mark)
        mark(?Player ?Game ?Mark)
        board(?Game ?Row1 ?Col ?Mark)
        mod1(?Row ?Row1)
        mod2(?Row ?Row2)
        excuse(unless(board(?Game ?Row2 ?Col ?Opponent-mark)))
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Try-column-next-to-next-own:
  Infer move(?Player ?Row ?Col ?Game)
  from (other-player(?Player ?Opponent ?Game)
        mark(?Opponent ?Game ?Opponent-mark)
        mark(?Player ?Game ?Mark)
        board(?Game ?Row2 ?Col ?Mark)
        mod1(?Row ?Row1)
```

```
mod2(?Row ?Row2)
        excuse(unless(board(?Game ?Row1 ?Col ?Opponent-mark)))
        excuse(unless(board(?Game ?Row ?Col ?Any-mark))))
Select-center:
  Infer move(?Player one one ?Game)
  from (excuse(unless(board(?Game one one ?Any-mark))))
Select-corner-00:
  Infer move(?Player zero zero ?Game)
  from (excuse(unless(board(?Game zero zero ?Any-mark))))
Select-corner-02:
  Infer move(?Player zero two ?Game)
  from (excuse(unless(board(?Game zero two ?Any-mark))))
Select-corner-20:
  Infer move(?Player two zero ?Game)
  from (excuse(unless(board(?Game two zero ?Any-mark))))
Select-corner-22:
  Infer move(?Player two two ?Game)
  from (excuse(unless(board(?Game two two ?Any-mark))))
Select-middle-01:
  Infer move(?Player zero one ?Game)
  from (excuse(unless(board(?Game zero one ?Any-mark))))
Select-middle-10:
  Infer move(?Player one zero ?Game)
  from (excuse(unless(board(?Game one zero ?Any-mark))))
Select-middle-12:
  Infer move(?Player one two ?Game)
  from (excuse(unless(board(?Game one two ?Any-mark))))
Select-middle-21:
                                                  MCC Carnot Project
```

Infer move(?Player two one ?Game)
from (excuse(unless(board(?Game two one ?Any-mark))))

assert-instance(game game1)
prefer-mark(PlayerB X)

#### A.4 Playing a Game

The game is played by creating three RAD agents, named Referee, PlayerA, and PlayerB, (arbitrarily) all connected to the same ESS. An interface for the human observer is created by

raduser -a referee -e delphi -u observer

which causes the response

RAD User Interface

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Connecting to ESS on Delphi+17001... ESS: Aide observer-aide is ready. Interface: pinging agent referee...

referee: READY

observer:

The appropriate knowledge bases should now be loaded into each agent by contacting them and issuing consult commands:

observer: consult "referee-kb.rad" referee: Consulting referee-kb.rad Finished consulting referee-kb.rad

observer: contact playera

playera: READY

observer: consult "playera-kb.rad" playera: Consulting playera-kb.rad

playera: Finished consulting playera-kb.rad

observer: contact playerb

playerb: READY

observer: consult "playerb-kb.rad" playerb: Consulting playerb-kb.rad

playerb: Finished consulting playerb-kb.rad

The Referee is now instructed to begin the game, and the results are shown to the Observer.

observer: contact referee

referee: READY

observer: play-tic-tac-toe(playera playerb game1) referee: Assertion PLAY-TIC-TAC-TOE-1 accepted.

Move 1: square 1 1 is occupied by PLAYERA
Move 2: square 1 1 is occupied by PLAYERB
Move 3: square 1 1 is occupied by PLAYERA
Move 4: square 1 1 is occupied by PLAYERB
Move 5: square 1 1 is occupied by PLAYERA
Move 6: square 1 1 is occupied by PLAYERB
Move 7: square 1 1 is occupied by PLAYERA
Move 8: square 1 1 is occupied by PLAYERA
Move 9: square 1 1 is occupied by PLAYERA

Game GAME1 ends in a tie. Game GAME1 is over.

## Appendix B

# Resolving Contradictions

### **B.1** Simple Example of CRM

The following is the contents of a file that RAD can consult for a simple example of contradiction resolution:

```
;;; Create class of elephants
assert-instance (class elephant)
;;; Make an attribute of color:
                                        ; Make color a relation
assert-instance (relation color)
argument-types(color (elephant symbol)); Specify domain, range
                                        ; Enable inheritance
specificity-ordered(color)
                                         ; Make it single-valued
single-valued(color)
                                         ; Provide default value
assert color(?X:elephant grey)
;;; Domain Rules, Assertions
;; Forward rule about albinoism
if albino(?X) then color(?X white)
if (says(?X ?Y) unless(liar(?X))) then assert(?Y)
Prefer(liar(?X) contradiction())
```

```
if (color(?X:elephant ?Y) bad-color(?X ?Y)) then contradiction()
;;; Situation-Specific Knowledge
assert-instance (elephant clyde)
says(joe albino(clyde))
RAD will exhibit the following behavior when the knowledge in this file is
consulted by an agent named advisor and accessed by a user named doctor:
doctor: consult "clyde.rad"
advisor: Consulting clyde.rad
advisor: Finished consulting clyde.rad
doctor: why color-3
advisor: Instance Slot Value COLOR-3
                                          (IN)
  COLOR(CLYDE WHITE)
was derived from the following:
                         (IN)
Forward Rule FRULE-0
  If
   ALBINO(?X)
  then
   COLOR(?X WHITE)
Assertion ALBINO-1
                       (IN)
   ALBINO(CLYDE)
doctor: why albino-1
 advisor: Assertion ALBINO-1
                                 (IN)
   ALBINO(CLYDE)
 was derived from the following:
                          (IN)
 Forward Rule FRULE-1
   If
```

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```
SAYS(?X ?Y)
  /UNLESS/0(?X PI::*ARG-END-SYM* LIAR)
  then
   ASSERT(?Y)
Assertion SAYS-1
                    (IN)
  SAYS(JOE ALBINO(CLYDE))
                      (TUO)
Failed Goal LIAR-1
  LIAR(JOE)
doctor: contradict color-3
advisor: CONTRADICTION-0
RESOLVED by justifying the following:
LIAR-2
doctor: why liar-2
advisor: Assertion LIAR-2
                              (IN)
  LIAR(JOE)
resolves contradictions:
CONTRADICTION-0
when it is justified by the following:
                         (IN)
Forward Rule FRULE-1
  If
   SAYS(?X ?Y)
   /UNLESS/0(?X PI::*ARG-END-SYM* LIAR)
  then
    ASSERT(?Y)
Assertion SAYS-1
   SAYS(JOE ALBINO(CLYDE))
```

```
Forward Rule FRULE-0
                        (IN)
  If
   ALBINO(?X)
  then
   COLOR(?X WHITE)
doctor:
        Complex Example of CRM
         Knowledge Base for Example of CRM
B.2.1
The following is the consult file for the example discussed in Chapter 8:
;;; Domain Rules, Assertions
assert-instance(relation
                test-error mistaken-observation result)
If Symptom(?Patient dehydrated)
then Conclusion(?Patient low amt H20)
If (Conclusion(?Patient low amt H20)
    Remember(Normal(?Patient amt Na)))
then Conclusion(?Patient high conc Na)
Infer Normal(?Patient amt Na)
from Unless(Result(?Patient low amt Na))
 If (Lab-test(?Patient low ?Meas ?Chem)
     Conclusion(?Patient high ?Meas ?Chem))
 then contradiction()
 ;;; Metaknowledge
 Fix(Lab-test(?Patient ?Degree ?Meas ?Chem)
     Lab-test(?Patient ?Degree ?Meas ?Chem)
     Test-error(?Patient ?Degree ?Meas ?Chem))
```

```
Fix(Conclusion(?Patient ?Degree ?Meas ?Chem)
    Symptom(?Patient ?Symptom)
    Mistaken-observation(?Patient ?Symptom))
```

Prefer(Mistaken-observation(?Patient ?Symptom) contradiction())
Prefer(Test-error(?Patient ?Degree ?Meas ?Chem) contradiction())
Prefer(Result(?Patient low amt ?Chem) contradiction())

#### **B.2.2** Protocol for Use of CRM

The following is the protocol using the consult file above. There are three agents. Two are radusers: the doctor and the lab. Agent "advisor" is a radagent. The consult file name is "sodium.rad".

doctor: contact advisor

advisor: READY

doctor: consult "sodium.rad" advisor: Consulting sodium.rad

advisor: Finished consulting sodium.rad

doctor: symptom(Jane dehydrated)

doctor: show conclusion

advisor: CONCLUSION Instance of:

RELATION

ARGUMENT-TYPES:

none

```
SPECIFICITY-ORDERED: no
   SINGLE-VALUED:
                        no
                        no
  PRIMITIVE:
   SYSTEM:
                        no
   PERSISTENT:
                        no
   UNIDIRECTIONAL:
                        none
   AXIOMATIC:
                        no
   SPECIALIZATION:
                        none
User Relation CONCLUSION (quaternary)
Current assertions:
CONCLUSION-1: CONCLUSION(JANE LOW AMT H20)
CONCLUSION-2: CONCLUSION (JANE HIGH CONC NA)
doctor: why conclusion-2
advisor: Assertion CONCLUSION-2
                                    (IN)
  CONCLUSION (JANE HIGH CONC NA)
was derived from the following:
Forward Rule FRULE-1
                         (IN)
  If
   CONCLUSION (?PATIENT LOW AMT H20)
   REMEMBER(NORMAL(?PATIENT AMT NA))
  )
  then
   CONCLUSION (?PATIENT HIGH CONC NA)
Assertion NORMAL-2
                       (IN)
  NORMAL (JANE AMT NA)
                           (IN)
Assertion CONCLUSION-1
  CONCLUSION (JANE LOW AMT H20)
```

At this point, the doctor has used the advisor as a simple expert system. By asserting a symptom, the rules have drawn conclusions which can be examined by the doctor. However, since the system is distributed, the lab may add its own data about the case.

lab: lab-test(Jane low conc Na)

advisor: CONTRADICTION-0

RESOLVED by justifying the following:

MISTAKEN-OBSERVATION-1

When the lab added conflicting information, a contradiction was produced. Because the consult file used did not include query schemata, no one was asked about this possibility. But the metaknowledge included was sufficient for the advisor to treat this as the most likely possibility and justify it.

However, this now comes to the attention of the doctor who disagrees that he may not have properly observed Jane's symptoms. This ability to disagree is an important part of RAD's functionality.<sup>1</sup>

doctor: show mistaken-observation-1

advisor: Assertion MISTAKEN-OBSERVATION-1 (IN)
MISTAKEN-OBSERVATION(JANE DEHYDRATED)

doctor: contradict mistaken-observation-1

advisor: CONTRADICTION-1

RESOLVED by justifying the following:

TEST-ERROR-1

The doctor finds this resolution likely. However, the lab does not believe this is a possibility and disagrees...

lab: show test-error-1

advisor: Assertion TEST-ERROR-1 (IN)

TEST-ERROR(JANE LOW CONC NA)

lab: contradict test-error-1

advisor: CONTRADICTION-2

RESOLVED by justifying the following:

RESULT-2

doctor: show result-2

<sup>&</sup>lt;sup>1</sup>The doctor may also have disagreed by commanding wrong after a query about mistaken observations.

advisor: Assertion RESULT-2 (IN)
RESULT(JANE LOW AMT NA)

The advisor has now offered another possibility: namely that Jane has an unusually low amount of sodium in her blood. Only this would explain why she was dehydrated but has a low blood concentration of sodium. This was a somewhat uncommon possibility discovered by backtracking. (The rule writer encoded this possibility as an assumption of normality.)

However, this result is still itself only an assumption. The doctor must still find out why Jane's sodium level is low. (It turns out she was radiation poisoned, which also accounts for the dehydration symptoms.) While pursuing this possibility, the doctor may want to remember the reasoning behind it. In the current version of RAD, this is done by a full replay of the justification for the assertion.

doctor: why result-2
advisor: Assertion RESULT-2 (IN)
 RESULT(JANE LOW AMT NA)

resolves contradictions:
CONTRADICTION-2
CONTRADICTION-1
CONTRADICTION-0

when it is justified by the following:

Backward Rule NORMAL-1 (IN)
 Infer
 NORMAL(?PATIENT AMT NA)
 from
 /UNLESS/0(?PATIENT LOW AMT NA PI::\*ARG-END-SYM\* RESULT)

```
(IN)
Forward Rule FRULE-0
  If
   SYMPTOM(?PATIENT DEHYDRATED)
   CONCLUSION (?PATIENT LOW AMT H20)
Forward Rule FRULE-1
                         (IN)
  If
   CONCLUSION (?PATIENT LOW AMT H20)
   REMEMBER (NORMAL (?PATIENT AMT NA))
  then
   CONCLUSION (?PATIENT HIGH CONC NA)
Forward Rule FRULE-2
                          (IN)
  If
   LAB-TEST (?PATIENT LOW ?MEAS ?CHEM)
   CONCLUSION (?PATIENT HIGH ?MEAS ?CHEM)
  )
  then
   CONTRADICTION()
doctor: quit
```

The information in this explanation amounts to the fact that two other possibilities, explaining conflicting data, were rejected and this was the remaining explanation. The doctor can see the other possibilities by commanding why on the contradictions listed. Eventually, a new justification, based on the discovery of radiation poisoning, can be established for this assertion.

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### Index

Contradiction 55 ! 59, 65 , 97 Count-proofs 59 = 52Cut 59, 65 Database Connection 24, 110, 115 ? 97, 108 Database Disconnect 24, 110, 115 ?? 108 Database Message 23 Accept 54, 103 DB-Connect 24, 110, 115 Actions 47, 50, 69 DB-Disconnect 24, 110, 115 Agenda 71 Aide 24 DB-Query 110 Default Values 42 And 52 Demand Message 22 Argument-types 47, 60 Do 57, 69 Ask 57, 112 Echo 114 Ask-once 58 Element 59 Assert 54, 69, 104 Eq 51 Assert-instance 54, 105 Assert-subclass 54, 105 Erase 55, 69, 106 Error Message 23 Atom 51 Atomic 51 Ess 92, 116 Axiomatic 46 Excuse 52 Explain 109 Backward Chaining 62, 65 Bagof 52, 73 Fail 54, 65 Fgrind 115 Bgrind 62, 115 Bottom 38 Forward Chaining 69 Frules-indexed 59 Check 101, 114 General Assertion 39 Clear 105 Ground 51 Comments 94 Consult 54, 69, 105 Halt-agent 116 Contact 92, 115 Help 116 Host 59 Contradict 106

Remove-subclass 55, 107 IN 27 Reply 107 Instance 32, 56 Response Message 23 Instance\* 32, 56 Said 53 Integerp 51 **Show 110** Is 56, 69, 99 Single-valued 48 Kill 55, 69, 106 Single-valued Relation 40, 42 Known 52, 73 Specialization 49 Lisp Functions 56, 97 Specificity-ordered 45, 49, 66 Lisp Macros 100 Stability 27 List-agents 92, 116 Statement Message 23 Location 92, 115, 116 Subclass 32, 56 Metaclasses 34, 60 Multiple-valued Relation 40 Subclass\* 32, 56 Supporters 27 Neq 51 **Next 109** System 47 T 38, 48 Nonvar 51 Tell 15, 55, 69, 107 Numberp 51 Toggle 108 Odd Loops 31 Translate 114 Orr 52 Triggerable 49 OUT 28 True 54 Particular Assertion 39 Types 37 Persistent 48 Unidirectional 49 Premise 39 Unless 53, 62, 100 Primitive 46 Unregister 24 Print 59, 69 Unreliable 59 Provable 52 Valid Justification 27 Prove 14, 52 Var 51 Prove-once 14, 53 Variables 37 Query 14, 53, 109 Warren Abstract Machine (WAM) Query-once 14, 53, 110 62 Quit 117 Was-told 53 Radagent 90 Well-foundedness 27 Raduser 91 Who 26 Register 24 Why 111 Reliable 55, 59 Reliable, 107 Wrong 106, 108

158

MCC Carnot Project

Remember 53, 107