Expert Systems Knowledge Based Systems

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Example Areas of Use

- Medical diagnosis
 - » Disease identification

Example Areas of Use – 2

- Medical diagnosis
 - » Disease identification
- Natural resource exploration
 - » Analyzing geological data

Example Areas of Use

- Medical diagnosis
 - » Disease identification
- Natural resource exploration
 - » Analyzing geological data
- Oustomizing complex equipment
 - » Computer systems

Properties

♦ Behaves like an expert in a narrow area

Properties – 2

- Sehaves like an expert in a narrow area
- ♦ Has a knowledge base of the information in the area

Properties – 3

- Sehaves like an expert in a narrow area
- ♦ Has a knowledge base of the information in the area
- ♦ Ability to explain its behaviour

Properties – 4

- Sehaves like an expert in a narrow area
- ♦ Has a knowledge base of the information in the area
- ♦ Ability to explain its behaviour
- ♦ Ability to deal with uncertain data

Structure



If ... then ... else ... rules

- ♦ The most popular form of knowledge representation
- Typical types of rules
 - » If condition P holds then conclude C
 - » If situation S exists then do action A
 - » If conditions P and Q hold then conditions C1 and C2 cannot hold

If ... then ... else ... examples

See Figures 15.2, 15.3 & 15.4 in Bratko

Kitchen leak example

- ♦ Figure 15.5 in Bratko
 - » How do you read the graph?

Kitchen leak example – 2

- ♦ Figure 15.5 in Bratko
 - » Can see how if...then...else rules can represent the graph on the left hand side
 - » Note the use of AND / OR for inputs
 - > Arc represents AND of inputs
 - > No arc represents OR of inputs

Properties of if...then...else rules

Moduarity

» Each rule or group of rules encapsulates a part of the domain

Properties of if...then...else rules – 2

- Moduarity
 - » Each rule or group of rules encapsulates a part of the domain
- ◊ Incrementability
 - » Add / delete rules as needed

Properties of if...then...else rules – 3

- Moduarity
 - » Each rule or group of rules encapsulates a part of the domain
- Incrementability
 - » Add / delete rules as needed
- ♦ Modifiability
 - » Can modify small parts of the knowledge as needed

Properties of if...then...else rules – 4

- Moduarity
 - » Each rule or group of rules encapsulates a part of the domain
- Incrementability
 - » Add / delete rules as needed
- ♦ Modifiability
 - » Can modify small parts of the knowledge as needed
- Supports transparency
 - » Relatively easy to explain and guide system's behaviour

Probabilistic behaviour

- Output Can extend rule syntax to include probability information
 - » If condition A then conclude C with probability P

» See Figure 15.3 in Bratko

Inference Engine

- » With if...then...else rules there are two ways of making inferences.
- **» What are they?**

Inference Engine – 2

- » With if...then...else rules there are two ways of making inferences.
- **» What are they?**
 - > Backward chaining
 - > Forward chaining

Backward chaining

» What is backward chaining?

Backward chaining – 2

» What is backward chaining?

> The way Prolog works from conclusions to the base facts, the confirmed facts, the evidence

> See pages 348..349

» What are the problems with backward chaining?

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 - > Syntax is not suitable of people unfamiliar with Prolog
 - > Cannot distinguish knowledge base from the rest of the system

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- **»** How can we overcome these problems?

- » What are the problems with backward chaining?
 - > Syntax is not suitable of people unfamiliar with Prolog
 - > Cannot distinguish knowledge base from the rest of the system
- **»** How can we overcome these problems?
 - > Customize the syntax with new operators
 - Bottom of page 349

» What do we need to do?

- » What do we need to do?
 - > Build an inference engine for the new rules
 - Figure 15.6 and program text

Forward chaining

» What is forward chaining?

Forward chaining – 2

» What is forward chaining?

> Work from the base facts, the confirmed facts, the evidence, to the conclusion

Forward chaining – 3

» What is forward chaining?

> Work from the base facts, the confirmed facts, the evidence, to the conclusion

» What do we need to do?

Forward chaining – 4

» What is forward chaining?

> Work from the base facts, the confirmed facts, the evidence, to the conclusion

- » What do we need to do?
 - > Build a forward chained inference engine
 - Figure 15.7 and program text

Forward vs backward chaining

Abstract view

Backward chaining <----input information → ... → derived information -----> Forward chaining

See Figure 15.5

Forward vs backward chaining – 2

- More concrete views
 - » data \rightarrow ... \rightarrow goals
 - » evidence \rightarrow ... \rightarrow hypotheses
 - » findings, observations $\rightarrow \dots \rightarrow$ explanations, diagnosis
 - » manifestations \rightarrow ... \rightarrow diagnoses, causes

Which is better

- ♦ Forward chaining?
- ♦ Backward chaining?

Which is better

Opends upon the problem
Which is better – 2

- Opends upon the problem
 - » Check if a hypothesis is true
 - > Work backward

Which is better – 3

- Opends upon the problem
 - » Check if a hypothesis is true
 - > Work backward
 - » Many hypotheses, cannot choose > Work forward

Which is better – 3

- Opends upon the problem
 - » Check if a hypothesis is true
 - > Work backward
 - » Many hypotheses, cannot choose > Work forward
- ♦ Forward is better when
 - » Accumulating evidence, data

Shape heuristic

- » When input information is sparse relative to derived information
- » Work forward or backward?

Shape heuristic – 2

- » When input information is sparse relative to derived information
- » Work forward or backward?
 - > Use forward chaining

Shape heuristic – 3

- » When input information is sparse relative to derived information
- » Work forward or backward?
 - > Use forward chaining
- » When input information is rich relative to derived information
- » Work forward or backward?

Shape heuristic – 5

- » When input information is sparse relative to derived information
- » Work forward or backward?
 - > Use forward chaining
- » When input information is rich relative to derived information
- » Work forward or backward?
 - > Use backward chaining

Reality

♦ Do we work forward or backward?

Reality – 2

- ♦ As tasks get more complex
 - » Better to interleave forward and backward chaining

 Doctor uses initial observations, evidence to form hypothesis

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 - **» Forward direction**

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 - **» Forward direction**
- Oursues most likely hypothesis

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- ♦ Can lead to gathering more evidence

- Doctor uses initial observations, evidence to form hypothesis
 - **» Forward direction**
- Oursues most likely hypothesis
 - » Backward direction to find if there is confirming evidence
- Can lead to gathering more evidence, need new hypothesis
 - » Forward direction, again

♦ Top page 353

Generating explanations

» There are two types of explanation What are they?

Generating explanations

» There are two types of explanation What are they?

> How

> Why

Generating explanations – how

- ♦ How did you find the answer?
 - » What do you present?

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 - » What do you present?
 - > Typically present a path trace

- ♦ How did you find the answer?
 - » What do you present?
 - > Typically present a path trace
 - » There is a problem in the kitchen, which was concluded from the hall being wet and the bathroom dry
 - > And
 - » No water came from the outside, which was concluded from the window being closed

Proof tree

- ♦ The how answer is to print out the proof tree
 - » Top page 354, Figure 15.8

> Given program text

> Compare with Figure 15.6

backward chained interpreter

Proof tree – 2

- ♦ The how answer is to print out the proof tree
 - » Top page 354, Figure 15.8
 - > Given program text
 - > Compare with Figure 15.6
 - backward chained interpreter
- ♦ The main work is printing the result in a readable format

- The why explanation is required during the reasoning process
 - » What do you present?

- The why explanation is required during the reasoning process
 - » What do you present?
 - > The system asks the user for information

- The why explanation is required during the reasoning process
 - » The system asks the user for information
 - » The user may ask why

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 - » The user may ask why
 - » The system then gives an explanation

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 - > Pages 354 .. 355

- The why explanation is required during the reasoning process
 - » The system asks the user for information
 - » The user may ask why
 - » The system then gives an explanation
 - > Pages 354 .. 355
 - > Figure 15.9 and program text

Output: Chapter 15 introduces an ad hoc way of dealing with probabilities

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 - » We will not look at these methods
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 - > Use Bayesian networks
 - > A sound and correct way of dealing with probability and uncertainty
 - > Modern approach

Semantic networks & frames

♦ Structure facts so as to elicit information
Semantic networks & frames – 2

- Structure facts so as to elicit information
- Introduce concepts that were adapted by object-oriented programming

Semantic networks & frames – 3

- Structure facts so as to elicit information
- Introduce concepts that were adapted by object-oriented programming
- Amounts to adopting a particular style of programming and organizing a program

Semantic networks & frames – 4

- Structure facts so as to elicit information
- Introduce concepts that were adapted by object-oriented programming
- Amounts to adopting a particular style of programming and organizing a program
 - » Requires discipline
 - > The programming environment does not directly support the style

Semantic networks

- Onsist of
 - » Entities

Semantic networks – 2

- Onsist of
 - » Entities
 - **» Relations between Entities**

Semantic networks – 3

- Onsist of
 - » Entities
 - **» Relations between Entities**
 - » Some similarity with Entity-Relation model in databases

Semantic networks – 4

- Onsists of
 - » Entities
 - **» Relations between Entities**
 - » Some similarity with Entity-Relation model in databases
 - » A graph representation is used

Example semantic network

Page 361 Bratko, E4



Semantic method representation

- ♦ The graph is represented in Prolog as a set of facts
 - » Functor a relation name
 - » Arguments are the entities at the head and tail of a relation



moving_method (kiwi , walk)
colour (kiwi , brown)
is_a ('Kim' , kiwi)
active_at (kiwi , night)

Inheritance representation

 Inheritance can be represented as a custom rule for each relationship

```
moving_method ( X , Method ) :-
    is_a ( X , SuperX ) ,
    moving_method ( SuperX , Method) .
```

Inheritance representation – 2

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Inheritance representation – 3

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```
moving_method ( X , Method ) :-
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```

- Our Cumbersome to use extensively
 - » Need a separate rule for each relation

Inheritance representation – 2

- A better way is to have a general rule for is_a based on facts
 - » Argument to fact is a compound term
 > relation_name (Arg1 , Arg2)

```
fact ( Fact ) :- Fact , ! .
fact ( Fact ) :-
    Fact =.. [ Relation , Arg1 , Arg2 ] ,
        is_a ( Arg1 , SuperArg ) ,
        SuperFact =.. [ Relation , SuperArg , Arg2 ] ,
        fact ( SuperFact ) .
```

Frames

♦ A frame is data structure whose components are slots

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- Slots have a name and a value

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FRAME: bird	
a_kind_of:	animal
moving_method:	fly
active_at:	daylight

- ♦ A frame is data structure whose components are slots
- Slots have a name and a value
 - » The value can be
 - > What?

- ♦ A frame is data structure whose components are slots
- Slots have a name and a value
 - » The value can be
 - > Simple values

- ♦ A frame is data structure whose components are slots
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 - » The value can be
 - > Simple values
 - > References to other frames
 - > Procedures to compute the slot value
 - > Unfilled

- ♦ A frame is data structure whose components are slots
- Slots have a name and a value
 - » The value can be
 - > Simple values
 - > References to other frames
 - > Procedures to compute the slot value
 - > Unfilled
 - How would they be filled?

- ♦ A frame is data structure whose components are slots
- Slots have a name and a value
 - » The value can be
 - > Simple values
 - > References to other frames
 - > Procedures to compute the slot value
 - > Unfilled
 - They are filled by inference

Frame representation in Prolog

- ♦ The frame name is the functor for a set of predicates.
- The arguments of the predicate are
 - » The slot name
 - » The slot value

bird (a_kind_of , animal).
bird (moving_method , fly).
bird (active_at , daylight).

FRAME: bird

a_kind_of:animalmoving_method:flyactive_at:daylight

Frame inheritance

Inheritance is shown by having a slot value being the name of a frame



Frame instance

A frame instance references the frame of which it is an instance



Frame instance – 2

- ♦ A frame instance references a frame
- On override slot values



Frame example

- ♦ Figure 15.12
- ♦ Program text pages 365 .. 366