

COMP219: Artificial Intelligence

Lecture 16: Forward and Backward Chaining

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Rule-Based System Architecture

- A collection of **rules**
- A collection of **facts**
- An **inference engine**
- We might want to
 - See what new facts can be *derived*
 - *Ask* whether a fact is implied by the knowledge base and facts already known

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Overview

- **Last time**
 - Introduced the reasons for explicit knowledge representation
 - Discussed properties of knowledge representation schemes
 - Introduced rules as a form of knowledge representation
- **Today**
 - Introduce algorithms for reasoning with rules
 - Discuss some of the problems of rule-based representations
- Learning outcome covered today:
Distinguish the characteristics, and advantages and disadvantages, of the major knowledge representation paradigms that have been used in AI, such as production rules, semantic networks, propositional logic and first-order logic;

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Control Schemes



- Given a set of rules, there are essentially two ways we can use them to generate new knowledge
 - **forward chaining**
 - starts with the facts, and sees what rules apply (and hence what should be done) given the facts
 - data driven
 - **backward chaining**
 - starts with something to find out, and looks for rules that will help in answering it
 - goal driven

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Fire Alarm Example



R1: IF hot AND smoky THEN fire
R2: IF alarm_beeps THEN smoky
R3: IF fire THEN sprinklers_on

F1: alarm_beeps [Given]
F2: hot [Given]

- We need to make the consequents **actions**

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Fire Alarm Example



R1: IF hot AND smoky THEN ADD fire
R2: IF alarm_beeps THEN ADD smoky
R3: IF fire THEN DO switch_sprinklers_on
ADD sprinklers_on

F1: alarm_beeps [Given]
F2: hot [Given]

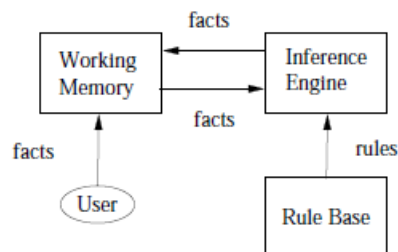
Forward Chaining

Use F1 and R2 to get F3 smoky
Use F2 and F3 and R1 to get F4 fire
Use F4 and R3 to get F5 sprinklers_on

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Forward Chaining

- In a forward chaining system
 - Facts are held in a **working memory**
 - Condition-action rules represent actions to take when specified facts occur in working memory
 - Often the actions involve adding or deleting facts from working memory



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Extending the Example



R1: IF hot AND smoky THEN ADD fire
R2: IF alarm_beeps THEN ADD smoky
R3: IF fire THEN DO switch_sprinklers_on
ADD sprinklers_on
R4: IF dry THEN DO switch_on_humidifier
ADD humidifier_on
R5: IF sprinklers_on THEN DELETE dry

F1: alarm_beeps; F2: hot; F3: dry

Now **two** rules match: R2 *and* R4

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Which rule to use?



- Use R2:
 - Add **smoky**: now R1 and R4 match
- Use R1:
 - Add **fire**: now R3 and R4 match
- Use R3:
 - Add **sprinklers_on**: R4 and R5 match
- Use R5:
 - Delete **dry**: now R4 does **not** match
- Note that R4 is **never** used in this sequence; so the choice **can affect the result**
- We have a **conflict**: we need a **conflict resolution strategy** to select the **right** rule

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Forward Chaining Algorithm

Repeat

Collect the rules whose conditions match facts in WM.

If more than one rule matches:

Use **conflict resolution strategy** to eliminate all but one.

Do actions indicated by the rules (add facts to WM or delete facts from WM).

Until problem is solved or no condition match.

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Conflict Resolution Strategy

- There are a number of approaches
 - Physically **order the rules**
 - hard to add rules to these systems
 - **Data** ordering
 - arrange problem elements in priority queue
 - use rule dealing with highest priority elements
 - **Specificity** or maximum specificity
 - based on number of conditions matching
 - choose the one with the most matches



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More Strategies

- **Recency** ordering
 - Data (based on order facts added to WM)
 - Rules (based on rule firings)
- **Context** limiting
 - partition rule base into disjoint subsets
 - we may order the subsets and we may also have preconditions
- **Random** selection
- Can also have combinations to break ties

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Meta Knowledge

- Another solution: use **meta-knowledge** (i.e. knowledge about knowledge) to guide search
- Example of meta-knowledge
IF
 conflict set contains any rule (c,a)
 such that a = ''animal is mammal''
THEN
 fire (c,a)
- So meta-knowledge encodes knowledge about how to guide search to solve the problem
- Explicitly coded in the form of rules, as with “object level” knowledge

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Application Areas

- Computer system configuration
 - Many possible set ups: forward chain from user needs
- Reactive robots
 - Get facts from environment and respond appropriately
- Conversational agents
 - Decide on the meaning of natural language input to give an appropriate response

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Properties of Forward Chaining



- Can be inefficient - lead to spurious rules firing, and unfocused problem solving (cf. breadth-first search)
- Set of rules that can fire known as **conflict set**
- Decision about which rule to fire - **conflict resolution**
- Different conflict resolutions may give different behaviour and different results

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Backward Chaining

- Same rules/facts may be processed differently, using backward chaining interpreter
- Backward chaining means reasoning from **goals back to facts**
- The idea is that this **focuses the search**
- Starts from a *goal* or *hypothesis*
 - Should I switch the sprinklers on?

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Backward Chaining Algorithm

To prove goal G:

If G is in the initial facts, it is proven.

Otherwise, find a rule which can be used to conclude G, and try to prove each of that rule's conditions (make conditions **sub-goals**).

- We add **goals**, not **facts** to working memory

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Using Prolog

- Prolog supports backward chaining directly:

```
alarm_beeps.
```

```
hot.
```

```
fire :- hot, smoky.
```

```
smoky :- alarm_beeps.
```

```
switch_on_sprinklers :- fire.
```

Conflict resolution is handled by clause order

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Fire Alarm Example



```
R1: IF hot AND smoky THEN ADD fire
```

```
R2: IF alarm_beeps THEN ADD smoky
```

```
R3: IF fire THEN DO switch_sprinklers_on  
      ADD sprinklers_on
```

```
F1: alarm_beeps;    F2: hot
```

- Goal: switch_sprinklers_on

Backward Chaining

R3 justifies goal if fire

R1 justifies fire if hot and smoky

Hot is a fact: R2 justifies smoky if alarm beeps

Alarm beeps is a fact

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Forward Chaining in Prolog

```
go(X):-member(sprinklers_on,X).
```

```
go(X):-member(fire,X), write([switching,sprinklers,on]),  
      go([sprinklers_on | X]).
```

```
go(X):-member(hot,X), member(smoky,X), go([fire | X]).
```

```
go(X):-member(alarm_beeps,X), go([smoky | X]).
```

```
?- go([hot,alarm_beeps]).
```

- **Argument acts as working memory**
- **Member succeeds if fact in working memory**
- **Conflict resolution through ordering of clauses**

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Exercise

Forward vs Backward Chaining

- Depends on problem, and on properties of rule set
- If you have clear hypotheses, backward chaining is likely to be better
 - Goal driven
 - Diagnostic problems or classification problems
 - Medical expert systems
- Forward chaining may be better if you have no clear hypothesis and want to see what can be concluded from current situation
 - Data driven
 - Synthesis systems
 - Configuration
 - Reactive systems

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Properties of Rules

- Rules are a natural representation
- They are inferentially adequate
- They are representationally adequate for some types of information/environments
- They can be inferentially inefficient (basically doing unconstrained search)
- They can have a well-defined syntax, but lack well-defined semantics
 - Conflict resolution can change their meaning

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Problems for Rules

- Inaccurate or incomplete information (inaccessible environments)
- Uncertain inference (non-deterministic environments)
- Non-discrete information (continuous environments)
- Default values
 - Anything that is not stated or derivable is false: they make the *closed world assumption*

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Summary

- We have looked at rules, which have often been used as a form of knowledge representation
- They can be used in either a **data** driven or a **goal** driven manner
 - Forward vs backward chaining
- **Next time**
 - We will look at a different form of knowledge representation: **structured objects**