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### **Todav**

See Russell and Norvig. chapter 7

- Propositional Logic ctd
- Inference algorithms

### Reminder

- **Syntax**: proposition symbols, joined with  $\neg$ ,  $\land$ ,  $\lor$ ,  $\Rightarrow$ ,  $\Leftrightarrow$ .
- **Semantics**: truth values, logical consequence  $KB \models F$
- special formulas:

valid:	true in all interpretations
satisfiable:	true in some interpretations
contradictory:	true in <b>no</b> interpretations



 $(\alpha \lor (\beta \land \gamma)) \equiv ((\alpha \lor \beta) \land (\alpha \lor \gamma))$  distributivity of  $\lor$  over  $\land$ 

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## **Proof methods**

Proof methods divide into (roughly) two kinds:

#### Application of inference rules

- Legitimate (sound) generation of new sentences from old
- Proof = a sequence of inference rule applications
   Can use inference rules as operators in a standard search alg.

#### Model checking

- truth table enumeration (always exponential in n) improved backtracking, heuristic search in model space
- (sound but incomplete)
- e.g., min-conflicts-like hill-climbing algorithms

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Idea: fire any rule whose premises are satisfied in the KB, add its conclusion to the KB, until query is found



# Forward and backward chaining

Horn Form (restricted)

 $\mathsf{KB}=\textit{conjunction}$  of Horn clauses also called definite clauses Horn clause =

- $\diamondsuit\,$  proposition symbol; or
- $\diamondsuit$  (conjunction of symbols)  $\Rightarrow$  symbol

 $\mathsf{E.g.},\ C \land (B \ \Rightarrow \ A) \land (C \land D \ \Rightarrow \ B)$ 

Modus Ponens (for Horn Form): complete for Horn KBs

$$\frac{\alpha_1,\ldots,\alpha_n,\qquad\alpha_1\wedge\cdots\wedge\alpha_n\Rightarrow\beta}{\beta}$$

Can be used with forward chaining or backward chaining. These algorithms are very natural and run in *linear* time

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### Forward chaining algorithm







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# **Proof of completeness**

FC derives every atomic sentence that is entailed by  $K\!B$ 

- 1. FC reaches a fixed point where no new atomic sentences are derived.
- 2. Consider the final state as a model m, assigning true/false to symbols.
- 3. Every clause in the original KB is true in m *Proof*: Suppose a clause  $a_1 \land \ldots \land a_k \Rightarrow b$  is false in m. Then  $a_1 \land \ldots \land a_k$  is true in m and b is false in m. Therefore the algorithm has not reached a fixed point!
- 4. Hence m is a model of KB.
- 5. If  $KB \models q$ , then q is true in *every* model of KB, including m.

### **Backward chaining**

Idea: work backwards from the query q: to prove q by BC, check if q is known already, or prove by BC all premises of some rule concluding q

Avoid loops: check if new subgoal is already on the goal stack

Avoid repeated work: check if new subgoal 1) has already been proved true, or 2) has already failed









### **Backward chaining example**



Pros and cons of propositional logic

Propositional logic allows partial/disjunctive/negated information

meaning of  $B_{1,1} \wedge P_{1,2}$  is derived from meaning of  $B_{1,1}$  and of  $P_{1,2}$ 

(unlike natural language, where meaning depends on context)

S Meaning in propositional logic is *context-independent* 

Propositional logic has very limited expressive power

E.g., cannot say "pits cause breezes in adjacent squares" except by writing one sentence for each square

Propositional logic is *declarative*:

pieces of syntax correspond to facts

Propositional logic is *compositional* :

(unlike natural language)

(unlike most data structures and databases)

#### Forward vs. backward chaining

FC is data-driven, cf. automatic, unconscious processing, e.g., object recognition, routine decisions

May do lots of work that is irrelevant to the goal

BC is goal-driven, appropriate for problem-solving,

e.g., Where are my keys? How do I get into a PhD programme?

Complexity of BC can be *much less* than linear in size of KB

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#### First-order logic

Whereas propositional logic assumes world contains *facts*, first-order logic (like natural language) assumes the world contains

- Objects: people, houses, numbers, theories, Jacques Chirac, colours, soduko games, wars, centuries . . .
- Relations: red, round, bogus, prime, multistoried, brother of, bigger than, inside, part of, has colour, occurred after, owns, comes between, . . .
- Functions: father of, best friend, second innings of, one more than, beginning of . . .

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## **Syntax of FOL: Basic elements**

Constants  $KingJohn, 2, UN, \ldots$ Brother,  $>, \ldots$ Predicates Sqrt,  $LeftLeqOf, \ldots$ Functions Variables  $x, y, a, b, \ldots$ Connectives  $\land \lor \neg \Rightarrow \Leftrightarrow$ Equality = ΥЭ Quantifiers

#### **Atomic sentences**

Atomic sentence =  $predicate(term_1, \dots, term_n)$ or  $term_1 = term_2$ 

> Term =  $function(term_1, \ldots, term_n)$ or constant or variable

Brother(KingJohn, RichardTheLionheart) E.g.,

> (Length(LeftLegOf(Richard)), Length(LeftLegOf(KingJohn))) For human consumption, often write > (X, Y) as X > Y.



### Universal quantification

 $\forall \langle variables \rangle \langle sentence \rangle$ 

Everyone at Berkeley is smart:  $\forall x \; At(x, Berkeley) \Rightarrow Smart(x)$ 

 $\forall x \ P$  is true in a model m iff P is true with x being each possible object in the model

Roughly speaking, equivalent to the conjunction of instantiations of P

 $At(KingJohn, Berkeley) \Rightarrow Smart(KingJohn)$  $\wedge$  At(Richard, Berkeley)  $\Rightarrow$  Smart(Richard)  $\wedge$  At(Berkeley, Berkeley)  $\Rightarrow$  Smart(Berkeley)  $\wedge \dots$ 

 $>(1,2) \lor <(1,2)$ 

 $>(1,2) \land (\neg >(1,2))$ 

Complex sentences are made from atomic sentences using connectives

 $\neg S$ ,  $S_1 \wedge S_2$ ,  $S_1 \vee S_2$ ,  $S_1 \Rightarrow S_2$ ,  $S_1 \Leftrightarrow S_2$ 

E.g.  $Sibling(KingJohn, Richard) \Rightarrow Sibling(Richard, KingJohn)$ 



is true if there is anyone who is not at Stanford!

