

CMSC 671 Fall 2010

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FOL / Inference in FOL

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Some material adopted from Hwee Tou Ng's course slides



Knowledge based agents and Logic (review/summary)



Declarative approach

- Knowledge-based agents follow a declarative approach to system building
 - In contrast a procedural approach encodes desired behaviors directly as program code.
 - Knowledge and Inference are separte
 - Inference is domain independent
 - ^D pit[2,2] or pit[3,1]
 - If wumpus[1,1] then not wumpus[2,2]

Representation, reasoning, and logic

- Logics are formal languages for representing information such that conclusions can be drawn
- A knowledge representation language is defined by:
 - its syntax, which specifies all valid sentences in the language
 - its semantics, which defines the "meaning" or truth of each sentence

Propositional logic



- Syntax the proposition symbols P₁, P₂ are sentences
 ¬S (negation), S₁ ∧ S₂ (conjunction), S₁ ∨ S₂ (disjunction), S₁ ⇒ S₂ (implication), S₁ ⇔ S₂ (biconditional)
- Semantics

 $\begin{array}{lll} \neg S & \text{is true iff } S \text{ is false} \\ S_1 \wedge S_2 & \text{is true iff } S_1 \text{ is true and} & S_2 \text{ is true} \\ S_1 \vee S_2 & \text{is true iff } S_1 \text{ is true or} & S_2 \text{ is true} \\ S_1 \Rightarrow S_2 & \text{is true iff} & S_1 \text{ is false or} & S_2 \text{ is true} \\ \text{i.e.,} & \text{is false iff} & S_1 \text{ is true and} & S_2 \text{ is true} \\ S_1 \Leftrightarrow S_2 & \text{is true iff} & S_1 \text{ is true and} & S_2 \text{ is false} \\ S_1 \Leftrightarrow S_2 & \text{is true iff} & S_1 \Rightarrow S_2 \text{ is true and} & S_2 \text{ is true} \\ \end{array}$



First-order logic

 $\bigcirc \bigcirc \bigcirc \bigcirc \bigcirc$

- First-order logic (FOL) models the world in terms of
 - **Objects**, which are things with individual identities
 - **Properties** of objects that distinguish them from other objects
 - Relations that hold among sets of objects
 - Functions, which are a subset of relations where there is only one "value" for any given "input"
- Examples:
 - Objects: Students, lectures, companies, cars ...
 - Relations: Brother-of, bigger-than, outside, part-of, has-color, occursafter, owns, visits, precedes, ...
 - Properties: blue, oval, even, large, ...
 - Functions: father-of, best-friend, second-half, one-more-than ...



First-order logic (2)

- Uses Terms for referring to Objects
 - a constant symbol, a variable symbol, or an n-place function of n terms
 - *e.g. John, x, LeftLeg(John)*
- Uses Predicate Symbols for referring to Relations
 - e.g. *Brother*
- Atomic Sentences state facts. e.g. Brother(John, Sam)
- Complex Sentences are formed from atomic sentences connected by the logical connectives: ¬P, P∨Q, P∧Q, P→Q, P→Q, P↔Q
- **Quantified sentences** add quantifiers \forall and \exists



First-order logic (3)



- A well-formed formula (wff) is a sentence containing no "free" variables. That is, all variables are "bound" by universal or existential quantifiers.
 - $(\forall x)P(x,y)$ has x bound as a universally quantified variable, but y is free.



Entailment and derivation

Entailment: KB |= Q

• The needle (Q) being in the haystack

 Q is entailed by KB if and only if the conclusion is true in every logically possible world in which all the premises in KB are true.

Derivation: KB |- Q

- Finding the needle (Q) in the haystack (KB)
- We can derive Q from KB if there is a proof consisting of a sequence of valid inference steps starting from the premises in KB and resulting in Q



Inference



- $KB \mid_i \alpha =$ sentence α can be derived from *KB* by procedure *i*
- Soundness: derivations produce only entailed sentences
- Completeness: derivations can produce all entailed sentences



Inference rules

- Logical inference is used to create new sentences that logically follow from a given set of predicate calculus sentences (KB).
- An inference rule is sound if every sentence X produced by an inference rule operating on a KB logically follows from the KB. (That is, the inference rule does not create any contradictions)
- An inference rule is complete if it is able to produce every expression that logically follows from (is entailed by) the KB. (Note the analogy to complete search algorithms.)



Logical Inference Chapter 9



Entailment for FOL is semidecidable

 Algorithms exist that say yes to every entailed sentence, but no algorithm exists that also says no to every nonentailed sentence.





Inference rules for quantifiers

- Applied to sentences with quantifiers to obtain sentences without quantifiers
- Universal instantiation
 - $\square \forall x P(x) :: P(A)$
- Existential instantiation
 - $\Box \exists x P(x) :: P(F) \qquad \leftarrow skolem \ constant \ F$



Universal instantiation (a.k.a. universal elimination)

• We can infer any sentence obtained by substituting a ground term



Universal instantiation (a.k.a. universal elimination)

- If (\forall x) P(x) is true, then P(C) is true, where
 C is *any* constant in the domain of x
- Example:
 - $-(\forall x) eats(Ziggy, x) \Rightarrow eats(Ziggy, IceCream)$
- The variable symbol can be replaced by any ground term, i.e., any constant symbol or function symbol applied to ground terms only

Existential instantiation (a.k.a. existential elimination)

The variable is replaced by a single new constant symbol that does not appear elsewhere in the knowledge base.



Existential instantiation (a.k.a. existential elimination)

- From $(\exists x) P(x)$ infer P(c)
- Example:
 - $(\exists x) eats(Ziggy, x) \rightarrow eats(Ziggy, Stuff)$
- Note that the variable is replaced by a brand-new constant not occurring in this or any other sentence in the KB
- Also known as skolemization; constant is a skolem constant
- We don't want to accidentally draw other inferences about it by introducing the constant
- Convenient to use this to reason about the unknown object, rather than constantly manipulating the existential quantifier

Inference in FOL



- Propositionalization
 - Use inference rules for quantifiers to convert the knowledge base from first order logic to propositional logic
- Use propositional inference
- Approach is slow, unless the domain is small



Inference in FOL (2)

- Forward Chaining Algorithm
 - derive the goal from the axioms
- Backward chaining Algorithm
 - start with the goal and attempt to resolve them working backwards
- Notice is the same idea as for propositional logic.
- We only need to take care of the quantifiers:
 Generalized Modus Ponens (GMP)





Automating FOL inference with Generalized Modus Ponens



Generalized Modus Ponens

- - Apply modus ponens reasoning to generalized rules
 - Combines And-Introduction, Universal-Elimination, and Modus Ponens
 - From P(c) and Q(c) and $(\forall x)(P(x) \land Q(x)) \rightarrow R(x)$ derive R(c)
 - General case: Given
 - atomic sentences P₁, P₂, ..., P_N
 - **implication sentence** $(Q_1 \land Q_2 \land ... \land Q_N) \rightarrow R$
 - Q₁, ..., Q_N and R are atomic sentences
 - **substitution** subst(θ , P_i) = subst(θ , Q_i) for i=1,...,N
 - Derive new sentence: subst(θ, R)
 - Substitutions
 - subst(θ , α) denotes the result of applying a set of substitutions defined by θ to the sentence α
 - A substitution list $\theta = \{v_1/t_1, v_2/t_2, ..., v_n/t_n\}$ means to replace all occurrences of variable symbol v_i by term t_i
 - Substitutions are made in left-to-right order in the list
 - subst({x/IceCream, y/Ziggy}, eats(y,x)) = eats(Ziggy, IceCream)

(GMP)

Generalized Modus Ponens

- $\forall x \operatorname{King}(x) \land \operatorname{Greedy}(x) \Longrightarrow \operatorname{Evil}(x)$
- $\forall y \text{ Greedy}(y)$
- King(John)
- •? Evil(John)
- Applying the substitution {x/John, y/John} we can infer Evil(John)



Generalized Modus Ponens

 If there is some substitution θ that makes each of the conjuncts of the premise of the implication identical to the sentences already in the knowledge base, then we can assert the conclusion of the implication, after applying θ.



Unification

- Unification is a "pattern-matching" procedure
 - Takes two atomic sentences, called literals, as input
 - Returns "Failure" if they do not match and a substitution list, θ, if they do
- That is, $unify(p,q) = \theta$ means $subst(\theta, p) = subst(\theta, q)$ for two atomic sentences, p and q
- θ is called the most general unifier (mgu)
- All variables in the given two literals are implicitly universally quantified
- To make literals match, replace (universally quantified) variables by terms

Unification algorithm

```
procedure unify(p, q, \theta)
     Scan p and q left-to-right and find the first corresponding
       terms where p and q "disagree" (i.e., p and q not equal)
     If there is no disagreement, return \theta (success!)
     Let r and s be the terms in p and q, respectively,
       where disagreement first occurs
     If variable(r) then {
       Let \theta = union(\theta, \{r/s\})
       Return unify(subst(\theta, p), subst(\theta, q), \theta)
     } else if variable(s) then {
       Let \theta = union(\theta, \{s/r\})
       Return unify(subst(\theta, p), subst(\theta, q), \theta)
     } else return "Failure"
   end
```



Unification: Remarks

- Unify is a linear-time algorithm that returns the most general unifier (mgu), i.e., the shortest-length substitution list that makes the two literals match.
- In general, there is not a unique minimum-length substitution list, but unify returns one of minimum length
- A variable can never be replaced by a term containing that variable
 - Example: x/f(x) is illegal.
- This "occurs check" should be done in the above pseudocode before making the recursive calls



Unification examples

Example:

- parents(x, father(x), mother(Bill))
- parents(Bill, father(Bill), y)
- {x/Bill, y/mother(Bill)}

• Example:

- parents(x, father(x), mother(Bill))
- parents(Bill, father(y), z)
- {x/Bill, y/Bill, z/mother(Bill)}

Example:

- parents(x, father(x), mother(Jane))
- parents(Bill, father(y), mother(y))
- Failure



Automated inference for FOL

- Automated inference using FOL is harder than PL
 - Variables can potentially take on an infinite number of possible values from their domains
 - Hence there are potentially an infinite number of ways to apply the universal instantiation rule of inference
- *Godel's Completeness Theorem* says that FOL entailment is only semidecidable
 - If a sentence is true given a set of axioms, there is a procedure that will determine this
 - If the sentence is false, then there is no guarantee that a procedure will ever determine this—i.e., it may never halt



Horn clauses

• A Horn clause is a sentence of the form:

 $(\forall x) P_1(x) \land P_2(x) \land \dots \land P_n(x) \to Q(x)$

where

• there are 0 or more P_i s and 0 or 1 Q

^D the P_is and Q are positive (i.e., non-negated) literals

- Equivalently: $P_1(x) \lor P_2(x) \ldots \lor P_n(x)$ where the P_i are all atomic and *at most one* of them is positive
- Prolog is based on Horn clauses
- Horn clauses represent a *subset* of the set of sentences representable in FOL



Horn clauses II

• Special cases • $P_1 \land P_2 \land \dots \land P_n \rightarrow Q$ • $P_1 \land P_2 \land \dots \land P_n \rightarrow false$ • true $\rightarrow Q$

These are not Horn clauses:
p(a) ∨ q(a)
(P ∧ Q) → (R ∨ S)



Forward chaining

- Proofs start with the given axioms/premises in KB, deriving new sentences using GMP until the goal/query sentence is derived
- This defines a forward-chaining inference procedure because it moves "forward" from the KB to the goal [eventually]
- Inference using GMP is complete for KBs containing only Horn clauses



Forward chaining example

• KB:

- □ allergies(X) \rightarrow sneeze(X)
- $\ \ \, \ \ \, \text{cat}(Y) \wedge allergic\text{-to-cats}(X) \rightarrow allergies(X)$
- □ cat(Felix)
- allergic-to-cats(Lise)
- Goal:
 - sneeze(Lise)



Forward chaining algorithm



```
FORWARD-CHAIN(KB, SUBST(\theta, conclusion))
else for each p' in KB such that UNIFY(p', SUBST(\theta, FIRST(premises))) = \theta_2 do
FIND-AND-INFER(KB, REST(premises), conclusion, COMPOSE(\theta, \theta_2))
end
```



Backward chaining

- Backward-chaining deduction using GMP is also complete for KBs containing only Horn clauses
- Proofs start with the goal query, find rules with that conclusion, and then prove each of the antecedents in the implication
- Keep going until you reach premises
- Avoid loops: check if new subgoal is already on the goal stack
- Avoid repeated work: check if new subgoal
 - Has already been proved true
 - Has already failed



Backward chaining example

• KB:

- □ allergies(X) \rightarrow sneeze(X)
- □ $cat(Y) \land allergic-to-cats(X) \rightarrow allergies(X)$
- □ cat(Felix)
- allergic-to-cats(Lise)
- Goal:
 - sneeze(Lise)


Backward chaining algorithm

```
function BACK-CHAIN(KB, q) returns a set of substitutions
   BACK-CHAIN-LIST(KB, [q], \{\})
function BACK-CHAIN-LIST(KB, glist, \theta) returns a set of substitutions
   inputs: KB, a knowledge base
             qlist, a list of conjuncts forming a query (\theta already applied)
            \theta, the current substitution
   static: answers, a set of substitutions, initially empty
   if qlist is empty then return \{\theta\}
   q \leftarrow \text{First}(qlist)
       for each q'_i in KB such that \theta_i \leftarrow \text{UNIFY}(q, q'_i) succeeds do
          Add COMFOSE(\theta, \theta_i) to answers
       end
       for each sentence (p_1 \land \ldots \land p_n \Rightarrow q'_i) in KB such that \theta_i \leftarrow U \land IFY(q, q'_i) succeeds do
          answers \leftarrow BACX-CHAIN-LIST(KB, SUBST(\theta_i, [p_1 \dots p_n]), COMPOSE(\theta, \theta_i)) \cup answers
       end
   return the union of BACK-CHAIN-LIST(KB, REST(qlist), \theta) for each \theta \in answers-
```

Forward vs. backward chaining

- FC is data-driven
 - 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
 - Where are my keys? How do I get to my next class?
 - Complexity of BC can be much less than linear in the size of the KB



Completeness of GMP

- GMP (using forward or backward chaining) is complete for KBs that contain only Horn clauses
- It is *not* complete for simple KBs that contain non-Horn clauses
- The following entail that S(A) is true:

$$(\forall x) P(x) \rightarrow Q(x)$$

$$(\forall \mathbf{x}) \neg \mathbf{P}(\mathbf{x}) \rightarrow \mathbf{R}(\mathbf{x})$$

$$(\forall x) Q(x) \rightarrow S(x)$$

 $(\forall x) \ R(x) \rightarrow S(x)$

- If we want to conclude S(A), with GMP we cannot, since the second one is not a Horn clause
- It is equivalent to $P(x) \lor R(x)$





Automating FOL inference with resolution



Resolution

- Resolution is a sound and complete inference procedure for FOL
- Main idea: Two clauses can be resolved if they contain complementary literals
- Reminder: Resolution rule for propositional logic:
 - $\square P_1 \lor P_2 \lor \ldots \lor P_n$
 - $\ \ \, \neg P_1 \lor Q_2 \lor ... \lor Q_m$
 - Resolvent: $P_2 \lor ... \lor P_n \lor Q_2 \lor ... \lor Q_m$
- Examples
 - P and $\neg P \lor Q$: derive Q (Modus Ponens)
 - $(\neg P \lor Q)$ and $(\neg Q \lor R)$: derive $\neg P \lor R$
 - P and \neg P : derive False [contradiction!]
 - $(P \lor Q)$ and $(\neg P \lor \neg Q)$: derive True

Resolution (2)



- Propositional literals are complementary if one if the negation of the other
- FOL literals are complementary if one **unifies** with the negation of the other



Resolution in first-order logic

- Given sentences
 - $\begin{array}{c} P_1 \lor \ldots \lor P_n \\ Q_1 \lor \ldots \lor Q_m \end{array}$
- in conjunctive normal form:
 - each P_i and Q_i is a literal, i.e., a positive or negated predicate symbol with its terms,
- if P_j and ¬Q_k unify with substitution list θ, then derive the resolvent sentence:

 $subst(\theta, P_1 \lor ... \lor P_{j\text{-}1} \lor P_{j\text{+}1} \ldots P_n \lor Q_1 \lor \ldots Q_{k\text{-}1} \lor Q_{k\text{+}1} \lor \ldots \lor Q_m)$

Example

- from clause
- and clause
- derive resolvent
- using

 $P(x, f(a)) \lor P(x, f(y)) \lor Q(y)$

 $\neg P(z, f(a)) \lor \neg Q(z)$

 $P(z, f(y)) \lor Q(y) \lor \neg Q(z)$ $\theta = \{x/z\}$



Resolution refutation

- Given a consistent set of axioms KB and goal sentence Q, show that KB = Q
- **Proof by contradiction:** Add ¬Q to KB and try to prove false.
 - i.e., (KB |- Q) \leftrightarrow (KB $\land \neg$ Q |- False)
- Resolution is refutation complete: it can establish that a given sentence Q is entailed by KB, but can't (in general) be used to generate all logical consequences of a set of sentences
- Also, it cannot be used to prove that Q is **not entailed** by KB.
- Resolution won't always give an answer since entailment is only semidecidable
 - And you can't just run two proofs in parallel, one trying to prove Q and the other trying to prove ¬Q, since KB might not entail either one

Refutation resolution proof

 \neg allergies(w) v sneeze(w) \neg cat(y) v \neg allergic-to-cats(z) \lor allergies(z) w/z \neg cat(y) v sneeze(z) \lor \neg allergic-to-cats(z) cat(Felix) y/Felix $sneeze(z) \vee \neg allergic-to-cats(z)$ allergic-to-cats(Lise) z/Lise sneeze(Lise) ¬sneeze(Lise) false negated query

tree

Required intermediate tasks

- How to convert FOL sentences to conjunctive normal form (a.k.a. CNF, clause form):
 normalization and skolemization
- How to unify two argument lists, i.e., how to find their most general unifier (mgu) θ:
 unification
- How to determine which two clauses in KB should be resolved next (among all resolvable pairs of clauses) : resolution (search) strategy





Converting to CNF



Converting sentences to CNF

1. Eliminate all \leftrightarrow connectives

 $(P \leftrightarrow Q) \Longrightarrow \ ((P \rightarrow Q) \land (Q \rightarrow P))$

2. Eliminate all \rightarrow connectives

 $(\mathbf{P} \to \mathbf{Q}) \Longrightarrow (\neg \mathbf{P} \lor \mathbf{Q})$

3. Reduce the scope of each negation symbol to a single predicate

$$\neg \neg P \Longrightarrow P$$
$$\neg (P \lor Q) \Longrightarrow \neg P \land \neg Q$$
$$\neg (P \land Q) \Longrightarrow \neg P \lor \neg Q$$
$$\neg (Q \land Q) \Longrightarrow (\exists x) \neg P$$
$$\neg (\exists x) P \Longrightarrow (\exists x) \neg P$$
$$\neg (\exists x) P \Longrightarrow (\forall x) \neg P$$

4. Standardize variables: rename all variables so that each quantifier has its own unique variable name

Converting sentences to clausal form Skolem constants and functions

5. Eliminate existential quantification by introducing Skolem constants/functions

 $(\exists x)P(x) \Rightarrow P(C)$

C is a Skolem constant (a brand-new constant symbol that is not used in any other sentence)

 $(\forall x)(\exists y)P(x,y) \Longrightarrow (\forall x)P(x, f(x))$

since \exists is within the scope of a universally quantified variable, use a **Skolem function f** to construct a new value that **depends on** the universally quantified variable

- f must be a brand-new function name not occurring in any other sentence in the KB.
- E.g., $(\forall x)(\exists y)$ loves $(x,y) \Rightarrow (\forall x)$ loves(x,f(x))

In this case, f(x) specifies the person that x loves

Converting sentences to clausal form

- 6. Remove universal quantifiers by (1) moving them all to the left end; (2) making the scope of each the entire sentence; and (3) dropping the "prefix" part $Ex: (\forall x)P(x) \Rightarrow P(x)$
- 7. Put into conjunctive normal form (conjunction of disjunctions) using distributive and associative laws $(P \land Q) \lor R \Rightarrow (P \lor R) \land (Q \lor R)$ $(P \lor Q) \lor R \Rightarrow (P \lor Q \lor R)$
- 8. Split conjuncts into separate clauses
- 9. Standardize variables so each clause contains only variable names that do not occur in any other clause

An example

$(\forall x)(P(x) \to ((\forall y)(P(y) \to P(f(x,y))) \land \neg(\forall y)(Q(x,y) \to P(y))))$

2. Eliminate \rightarrow

 $(\forall x)(\neg P(x) \lor ((\forall y)(\neg P(y) \lor P(f(x,y))) \land \neg(\forall y)(\neg Q(x,y) \lor P(y))))$

3. Reduce scope of negation

 $(\forall x)(\neg P(x) \lor ((\forall y)(\neg P(y) \lor P(f(x,y))) \land (\exists y)(Q(x,y) \land \neg P(y))))$

4. Standardize variables

 $(\forall x)(\neg P(x) \lor ((\forall y)(\neg P(y) \lor P(f(x,y))) \land (\exists z)(Q(x,z) \land \neg P(z))))$

5. Eliminate existential quantification

 $(\forall x)(\neg P(x) \lor ((\forall y)(\neg P(y) \lor P(f(x,y))) \land (Q(x,g(x)) \land \neg P(g(x)))))$

6. Drop universal quantification symbols

 $(\neg P(x) \lor ((\neg P(y) \lor P(f(x,y))) \land (Q(x,g(x)) \land \neg P(g(x)))))$



Example



7. Convert to conjunction of disjunctions $(\neg P(x) \lor \neg P(y) \lor P(f(x,y))) \land (\neg P(x) \lor Q(x,g(x))) \land$ $(\neg P(x) \lor \neg P(g(x)))$

- 8. Create separate clauses
 - $\neg P(x) \lor \neg P(y) \lor P(f(x,y))$
 - $\neg P(x) \lor Q(x,g(x))$
 - $\neg P(x) \lor \neg P(g(x))$
- 9. Standardize variables $\neg P(x) \lor \neg P(y) \lor P(f(x,y))$ $\neg P(z) \lor Q(z,g(z))$ $\neg P(w) \lor \neg P(g(w))$





Unification

(see slides 25-28)





Resolution example



Practice example *Did Curiosity kill the cat?*

- Jack owns a dog. Every dog owner is an animal lover. No animal lover kills an animal. Either Jack or Curiosity killed the cat, who is named Tuna. Did Curiosity kill the cat?
- These can be represented as follows:
 - A. $(\exists x) \text{ Dog}(x) \land \text{Owns}(\text{Jack}, x)$
 - B. $(\forall x) ((\exists y) \text{Dog}(y) \land \text{Owns}(x, y)) \rightarrow \text{AnimalLover}(x)$
 - C. $(\forall x)$ AnimalLover $(x) \rightarrow ((\forall y) \text{ Animal}(y) \rightarrow \neg \text{Kills}(x,y))$
 - D. Kills(Jack,Tuna) v Kills(Curiosity,Tuna)
 - E. Cat(Tuna)
 - F. $(\forall x)$ Cat $(x) \rightarrow$ Animal(x)
 - G. Kills(Curiosity, Tuna)

- GOAL



Convert to clause form

- A1. (Dog(D)) D is a skolem constant
- A2. (Owns(Jack,D))
- B. $(\neg Dog(y), \neg Owns(x, y), AnimalLover(x))$
- C. (¬AnimalLover(a), ¬Animal(b), ¬Kills(a,b))
- D. (Kills(Jack,Tuna), Kills(Curiosity,Tuna))
- E. Cat(Tuna)
- F. $(\neg Cat(z), Animal(z))$
- Add the negation of query:

¬G: (¬Kills(Curiosity, Tuna))

The resolution refutation proof	
R1: ¬G, D, { }	(Kills(Jack, Tuna))
R2: R1, C, {a/Jack, b/Tuna}	(~AnimalLover(Jack), ~Animal(Tuna))
R3: R2, B, {x/Jack}	(~Dog(y), ~Owns(Jack, y), ~Animal(Tuna))
R4: R3, A1, {y/D}	(~Owns(Jack, D), ~Animal(Tuna))
R5: R4, A2, { }	(~Animal(Tuna))
R6: R5, F, {z/Tuna}	(~Cat(Tuna))
R7: R6, E, { }	FALSE





Resolution search strategies

- Repeated applications of the resolution inference rule will eventually find a proof if one exists.
- Some strategies help to find proofs efficiently



Resolution TP as search

- Resolution can be thought of as the bottom-up construction of a search tree, where the leaves are the clauses produced by KB and the negation of the goal
- When a pair of clauses generates a new resolvent clause, add a new node to the tree with arcs directed from the resolvent to the two parent clauses
- **Resolution succeeds** when a node containing the **False** clause is produced, becoming the **root node** of the tree
- A strategy is **complete** if its use guarantees that the empty clause (i.e., false) can be derived whenever it is entailed

Strategies



- There are a number of general (domain-independent) strategies that are useful in controlling a resolution theorem prover:
 - Unit preference
 - Set of support
 - Input resolution
 - Subsumption
 - Ordered resolution



Example



- **1.** ¬Battery-OK ∨ ¬Bulbs-OK ∨ Headlights-Work
- 2. ¬Battery-OK v ¬Starter-OK v Empty-Gas-Tank v Engine-Starts
- **3.** \neg Engine-Starts \lor Flat-Tire \lor Car-OK
- 4. Headlights-Work
- 5. Battery-OK
- 6. Starter-OK
- 7. ¬Empty-Gas-Tank
- 8. ¬Car-OK
- 9. ¬Flat-Tire

negated goal



Breadth-first search

- Level 0 clauses are the original axioms and the negation of the goal
- Level k clauses are the resolvents computed from two clauses, one of which must be from level k-1 and the other from any earlier level
- Compute all possible level 1 clauses, then all possible level 2 clauses, etc.
- Complete, but very inefficient



BFS example

- **1.** ¬Battery-OK ∨ ¬Bulbs-OK ∨ Headlights-Work
- 2. ¬Battery-OK v ¬Starter-OK v Empty-Gas-Tank v Engine-Starts
- 3. ¬Engine-Starts ∨ Flat-Tire ∨ Car-OK
- 4. Headlights-Work
- 5. Battery-OK
- 6. Starter-OK
- 7. ¬Empty-Gas-Tank
- **8.** ¬Car-OK
- 9. ¬Flat-Tire
- 1,4 10. ¬Battery-OK ∨ ¬Bulbs-OK
- 1,5 11. \neg Bulbs-OK \lor Headlights-Work
- 2,3 12. \neg Battery-OK $\lor \neg$ Starter-OK \lor Empty-Gas-Tank \lor Flat-Tire \lor Car-OK
- 2,5 13. \neg Starter-OK \lor Empty-Gas-Tank \lor Engine-Starts
- 2,6 14. \neg Battery-OK \lor Empty-Gas-Tank \lor Engine-Starts
- 2,7 15. \neg Battery-OK \neg Starter-OK \lor Engine-Starts
 - 16. ... [and we're still only at Level 1!]

Unit preference (unit resolution)

Shortest-clause heuristic:

Generate a clause with the fewest literals first

Unit resolution:

Prefer resolution steps in which at least one parent clause is a "unit clause," i.e., a clause containing a single literal

Not complete in general, but complete for Horn clause KBs



Unit resolution example

- **1.** \neg Battery-OK $\lor \neg$ Bulbs-OK \lor Headlights-Work
- 2. ¬Battery-OK v ¬Starter-OK v Empty-Gas-Tank v Engine-Starts
- 3. ¬Engine-Starts ∨ Flat-Tire ∨ Car-OK
- 4. Headlights-Work
- 5. Battery-OK
- 6. Starter-OK
- **7.** ¬Empty-Gas-Tank
- **8.** ¬Car-OK
- 9. ¬Flat-Tire
- 1,5 10. \neg Bulbs-OK \lor Headlights-Work
- 2,5 11. \neg Starter-OK \lor Empty-Gas-Tank \lor Engine-Starts
- 2,6 12. ¬Battery-OK ∨ Empty-Gas-Tank ∨ Engine-Starts
- 2,7 13. ¬Battery-OK ¬ Starter-OK ∨ Engine-Starts
- 3,8 14. ¬Engine-Starts ∨ Flat-Tire
- 3,9 15. ¬Engine-Starts ¬ Car-OK
 - 16. ... [this doesn't seem to be headed anywhere either!]



Set of support

- At least one parent clause must be the negation of the goal *or* a "descendant" of such a goal clause (i.e., derived from a goal clause)
- (When there's a choice, take the most recent descendant)
- Complete (assuming all possible set-ofsupport clauses are derived)
- Gives a goal-directed character to the search

Set of support example

- **1.** \neg Battery-OK $\lor \neg$ Bulbs-OK \lor Headlights-Work
- 2. ¬Battery-OK v ¬Starter-OK v Empty-Gas-Tank v Engine-Starts
- 3. ¬Engine-Starts ∨ Flat-Tire ∨ Car-OK
- 4. Headlights-Work
- 5. Battery-OK
- 6. Starter-OK
- 7. ¬Empty-Gas-Tank
- **8.** ¬Car-OK
- 9. ¬Flat-Tire
- 9,3 10. ¬Engine-Starts ∨ Car-OK
- 10,2 11. \neg Battery-OK $\lor \neg$ Starter-OK \lor Empty-Gas-Tank \lor Car-OK
- 10,8 12. –Engine-Starts
- 11,5 13. \neg Starter-OK \lor Empty-Gas-Tank \lor Car-OK
- 11,6 14. \neg Battery-OK \lor Empty-Gas-Tank \lor Car-OK
- 11,7 15. \neg Battery-OK $\lor \neg$ Starter-OK \lor Car-OK
 - 16. ... [a bit more focused, but we still seem to be wandering]



Unit resolution + set of support



example

- **1.** ¬Battery-OK ∨ ¬Bulbs-OK ∨ Headlights-Work
- 2. ¬Battery-OK v ¬Starter-OK v Empty-Gas-Tank v Engine-Starts
- 3. ¬Engine-Starts ∨ Flat-Tire ∨ Car-OK
- 4. Headlights-Work
- 5. Battery-OK
- 6. Starter-OK
- 7. ¬Empty-Gas-Tank
- **8.** ¬Car-OK
- 9. ¬Flat-Tire
- 9,3 10. ¬Engine-Starts ∨ Car-OK
- 10,8 11. –Engine-Starts
- 12,2 12. \neg Battery-OK $\lor \neg$ Starter-OK \lor Empty-Gas-Tank
- 12,5 13. \neg Starter-OK \lor Empty-Gas-Tank
- 13,6 14. Empty-Gas-Tank
- 14,7 15. FALSE

[Hooray! Now that's more like it!]



Simplification heuristics

• Subsumption:

Eliminate all sentences that are subsumed by (more specific than) an existing sentence to keep the KB small

- If P(x) is already in the KB, adding P(A) makes no sense P(x) is a superset of P(A)
- ^D Likewise adding $P(A) \lor Q(B)$ would add nothing to the KB

Tautology:

Remove any clause containing two complementary literals (tautology)

Pure symbol:

If a symbol always appears with the same "sign," remove all the clauses that contain it

 Equivalent to assuming that symbol to be always-true or alwaysfalse

(: can't draw any inferences about other symbols in the clause)

Example (Pure Symbol)



- 1. Battery OK Bulbs OK Headlights Work-
- 2. \neg Battery-OK $\lor \neg$ Starter-OK \lor Empty-Gas-Tank \lor Engine-Starts
- **3.** \neg Engine-Starts \lor Flat-Tire \lor Car-OK
- 4. Headlights Work
- 5. Battery-OK
- 6. Starter-OK
- 7. Empty-Gas-Tank
- **8.** *¬*Car-OK
- 9. ¬Flat-Tire


Input resolution

- At least one parent must be one of the input sentences (i.e., either a sentence in the original KB or the negation of the goal)
- Not complete in general, but complete for Horn clause KBs

Linear resolution

- Extension of input resolution
- One of the parent sentences must be an input sentence *or* an ancestor of the other sentence
- Complete



Ordered resolution

- Search for resolvable sentences in order (left to right)
- This is how Prolog operates
- Resolve the first element in the sentence first
- This forces the user to define what is important in generating the "code"
- The way the sentences are written controls the resolution





Using FOL



Using FOL

 $\bigcirc \bigcirc \bigcirc \bigcirc \bigcirc$

- **Domain M:** the set of all objects in the world (of interest)
- Assertions: sentences added to KB by using TELL (as in propositional logic)
 - TELL (KB, Person(Richard))
- **Queries/Goals:** ask questions of the KB. Any query that is logically entailed by the knowledge base should be answered affirmatively.
 - ASK (KB, Person(Richard))



Using FOL (2)

 $\bigcirc \bigcirc \bigcirc \bigcirc \bigcirc$

Can be cumbersome and prone to mistakes

Database systems approach

- Unique names assumption
- Closed-World assumption (not unknown truth)
- Domain closure



Using forward chaining

- Systems implementing forward chaining are known as production systems.
 - Perform efficient updates with very large rule sets
 - Example: Datalog



Using backward chaining

- Backward chaining is used in logic programming systems.
 - Employ sophisticated compiler technology to provide fast inference.
 - It is the mechanism underlying the execution of Prolog programs



Forward vs. backward chaining

- FC is data-driven
 - 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
 - Where are my keys? How do I get to my next class?
 - Complexity of BC can be much less than linear in the size of the KB



Prolog

- A logic programming language based on Horn clauses
 - Resolution refutation
 - Control strategy: goal-directed and depth-first
 - always start from the goal clause
 - always use the new resolvent as one of the parent clauses for resolution
 - backtracking when the current thread fails
 - complete for Horn clause KB
 - Support answer extraction (can request single or all answers)
 - Orders the clauses and literals within a clause to resolve non-determinism
 - Q(a) may match both $Q(x) \le P(x)$ and $Q(y) \le R(y)$
 - A (sub)goal clause may contain more than one literals, i.e., <= P1(a), P2(a)
 - Use "closed world" assumption (negation as failure)
 - If it fails to derive P(a), then assume ~P(a)

Prolog (2)



- Proof method of Prolog is resolution refutation
- Backward reasoning with sub-goaling: we assert the negated goal and try to work backwards, unifying and resolving clauses until we get to the empty clause



Summary



- Logical agents apply inference to a knowledge base to derive new information and make decisions
- Basic concepts of logic:
 - Syntax: formal structure of sentences
 - Semantics: truth of sentences wrt models
 - Entailment: necessary truth of one sentence given another
 - Inference: deriving sentences from other sentences
 - Soundness: derivations produce only entailed sentences
 - Completeness: derivations can produce all entailed sentences
- FC and BC are linear time, complete for Horn clauses
- Resolution is a sound and complete inference method for propositional and first-order logic

