CS 2740 Knowledge Representation Lecture 6

Propositional logic. Horn clauses

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Logical inference problem

Logical inference problem:

- Given:
 - a knowledge base KB (a set of sentences) and
 - a sentence α (called a theorem),
- Does a KB semantically entail α ? $KB \models \alpha$?

In other words: In all interpretations in which sentences in the KB are true, is also α true?

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Solving logical inference problem

In the following:

How to design the procedure that answers:

$$KB = \alpha$$
?

Three approaches:

- Truth-table approach
- Inference rules
- Conversion to the inverse SAT problem
 - Resolution-refutation

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KB in restricted forms

If the sentences in the KB are restricted to some special forms some of the sound inference rules may become complete

Example:

• Horn form (Horn normal form)

$$(A \lor \neg B) \land (\neg A \lor \neg C \lor D)$$

Can be written also as: (B

- $(B \Rightarrow A) \land ((A \land C) \Rightarrow D)$
- Two inference rules that are sound and <u>complete with</u> <u>respect to propositional symbols</u> for KBs in the Horn normal form:
 - Resolution (positive unit resolution)
 - Modus ponens

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KB in Horn form

Horn form: a clause with at most one positive literal

$$(A \lor \neg B) \land (\neg A \lor \neg C \lor D)$$

- Not all sentences in propositional logic can be converted into the Horn form
- KB in Horn normal form:
 - Two types of propositional statements:

• Rules
$$(\neg B_1 \lor \neg B_2 \lor \dots \neg B_k \lor A)$$

 $(\neg (B_1 \land B_2 \land \dots B_k) \lor A)$
 $(B_1 \land B_2 \land \dots B_k \Rightarrow A)$

• Propositional symbols: **facts** B

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KB in Horn form

- Application of the modus ponens:
 - Infers new facts from previous facts

$$\frac{B \Rightarrow A, \quad B}{A}$$

$$\underbrace{(B_1 \land B_2 \land \dots B_k \Rightarrow A), B_1, B_2, \dots B_k}_{A}$$

 Modus ponens is sound and complete with respect to propositional symbols for the KBs in the Horn normal form

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Question:

How efficient the inferences in HNF can be?

Answer:

Procedures linear in the size of the KB in the Horn form exist.

- Size of a clause: the number of literals it contains.
- Size of the KB in the HNF: the sum of the sizes of its elements.

Example:

$$A, B, (A \land B \Rightarrow C), (C \Rightarrow D), (C \Rightarrow E), (E \land F \Rightarrow G)$$
 or

$$A, B, (\neg A \lor \neg B \lor C), (\neg C \lor D), (\neg C \lor E), (\neg E \lor \neg F \lor G)$$

The size is: 12

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Complexity of inferences for KBs in HNF

How to do the inference? If the HNF (is in clausal form) we can apply resolution.

$$A, B, (\neg A \lor \neg B \lor C), (\neg C \lor D), (\neg C \lor E), (\neg E \lor \neg F \lor G)$$

$$-B \lor C$$

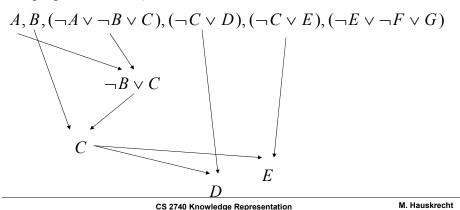
$$C$$

$$E$$

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Features:

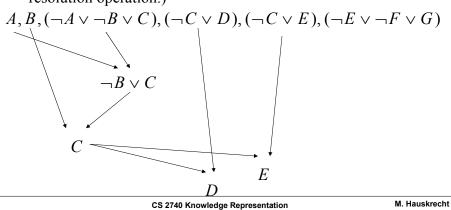
• Every resolution is a **positive unit resolution**; that is, a resolution in which **one clause is a positive unit clause** (i.e., a proposition letter).



Complexity of inferences for KBs in HNF

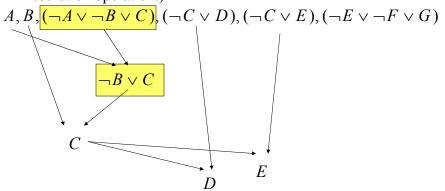
Features:

• At each resolution, the input clause which is not a unit clause is a logical consequence of the result of the resolution. (Thus, the input clause may be deleted upon completion of the resolution operation.)



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• At each resolution, the input clause which is not a unit clause is a logical consequence of the result of the resolution. (Thus, the input clause may be deleted upon completion of the resolution operation.)



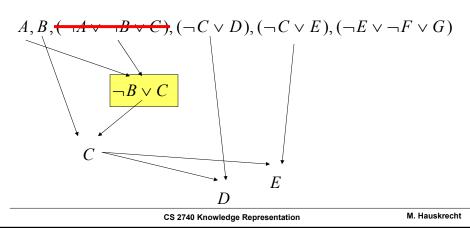
Complexity of inferences for KBs in HNF

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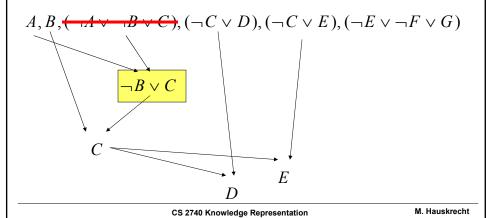
Features:

• Following this deletion, the size of the KB (the sum of the lengths of the remaining clauses) is one less than it was before the operation.)



Features:

• If n is the size of the KB, then at most n positive unit resolutions may be performed on it.



Complexity of inferences for KBs in HNF

Linear time algorithm:

- The number of positive unit resolutions is limited to the size of the formula (n)
- But to assure overall linear time we need to access each proposition in a constant time:
- Data structures indexed by proposition names may be accessed in constant time. (This is possible if the proposition names are number in a range (e.g., 1..n), so that array lookup is the access operation.
- If propositions are accessed by name, then a symbol table is necessary, and the algorithm will run in time $O(n \cdot log(n))$.

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Forward and backward chaining

Two inference procedures based on **modus ponens** for **Horn KBs**:

Forward chaining

Idea: Whenever the premises of a rule are satisfied, infer the conclusion. Continue with rules that became satisfied.

• Backward chaining (goal reduction)

Idea: To prove the fact that appears in the conclusion of a rule prove the premises of the rule. Continue recursively.

Both procedures are complete for KBs in the Horn form !!!

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Forward chaining example

Forward chaining

Idea: Whenever the premises of a rule are satisfied, infer the conclusion. Continue with rules that became satisfied.

Assume the KB with the following rules and facts:

KB: R1: $A \wedge B \Rightarrow C$

R2: $C \wedge D \Rightarrow E$

R3: $C \wedge F \Rightarrow G$

F1: A F2: B F3: D

Theorem: E?

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Forward chaining example

Theorem: E

KB: R1: $A \wedge B \Rightarrow C$

R2: $C \wedge D \Rightarrow E$

R3: $C \wedge F \Rightarrow G$

F1: A

F2: *B*

F3: *D*

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Forward chaining example

Theorem: E

KB: R1: $A \wedge B \Rightarrow C$

R2: $C \wedge D \Rightarrow E$

R3· $C \wedge F \Rightarrow G$

F1: A

F2: *B*

F3: *D*

Rule R1 is satisfied.

F4: *C*

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Forward chaining example

Theorem: E

KB: R1: $A \wedge B \Rightarrow C$

R2: $C \wedge D \Rightarrow E$

R3: $C \wedge F \Rightarrow G$

F1: A

F2: *B*

F3: *D*

Rule R1 is satisfied.

F4: *C*

Rule R2 is satisfied.

F5: *E*



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Forward chaining

- Efficient implementation: linear in the size of the KB
- Example:

$$P \Rightarrow Q$$

$$L \wedge M \Rightarrow P$$

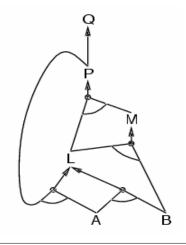
$$B \wedge L \Longrightarrow M$$

$$A \wedge P \Longrightarrow L$$

$$A \wedge B \Rightarrow L$$

 \boldsymbol{A}

В



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Forward chaining

• Runs in time linear in the number of literals in the Horn formulae

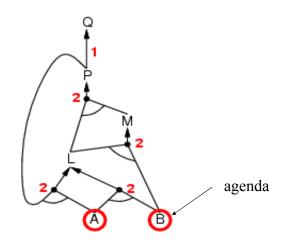
```
function PL-FC-Entails?(KB,q) returns true or false
local variables: count, a table, indexed by clause, initially the number of premises inferred, a table, indexed by symbol, each entry initially false agenda, a list of symbols, initially the symbols known to be true while agenda is not empty do p \leftarrow \text{POP}(agenda) unless inferred[p] do inferred[p] \leftarrow true for each Horn clause c in whose premise p appears do decrement count[c] if count[c] = 0 then do if \text{HEAD}[c] = q then return true \text{PUSH}(\text{HEAD}[c], agenda) return false
```

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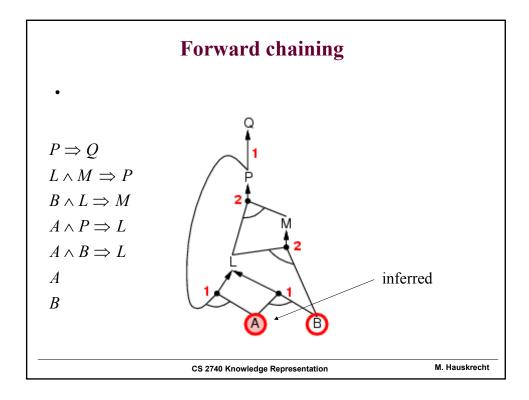
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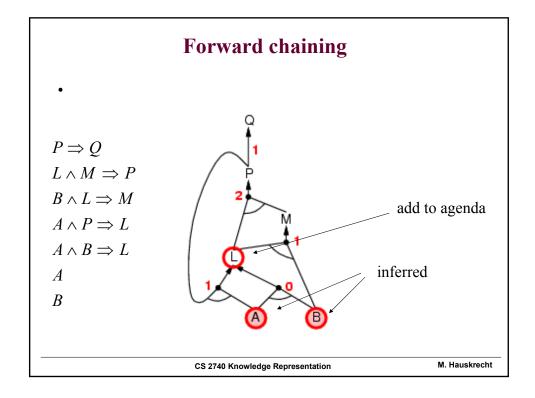
Forward chaining

 $P \Rightarrow Q$ $L \land M \Rightarrow P$ $B \land L \Rightarrow M$ $A \land P \Rightarrow L$ $A \land B \Rightarrow L$ A



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•

B

$$P \Rightarrow Q$$

$$L \land M \Rightarrow P$$

$$B \land L \Rightarrow M$$

$$A \land P \Rightarrow L$$

$$A \land B \Rightarrow L$$

$$A$$

A B

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Forward chaining

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$$P \Rightarrow Q$$

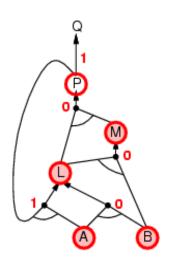
$$L \land M \Rightarrow P$$

$$B \land L \Rightarrow M$$

$$A \land P \Rightarrow L$$

$$A \land B \Rightarrow L$$

$$A$$



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Forward chaining

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$$P \Rightarrow Q$$

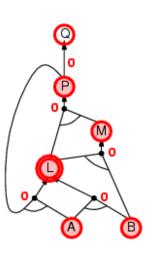
$$L \land M \Rightarrow P$$

$$B \land L \Rightarrow M$$

$$A \land P \Rightarrow L$$

$$A \land B \Rightarrow L$$

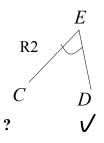
$$A$$



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Backward chaining example



KB: R1: $A \wedge B \Rightarrow C$

R2: $C \wedge D \Rightarrow E$

R3: $C \wedge F \Rightarrow G$

F1: A

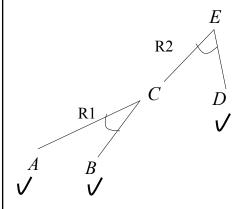
F2: *B*

F3: *D*

- Backward chaining is more focused:
 - tries to prove the theorem only

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Backward chaining example



- KB: R1: $A \wedge B \Rightarrow C$
 - R2: $C \wedge D \Rightarrow E$
 - R3: $C \wedge F \Rightarrow G$
 - F1: A
 - F2: *B*
 - F3: *D*

- Backward chaining is more focused:
 - tries to prove the theorem only

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Backward chaining

$$P \Rightarrow Q$$

$$L \wedge M \Rightarrow P$$

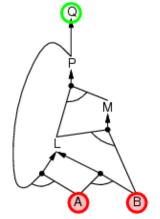
$$B \wedge L \Longrightarrow M$$

$$A \wedge P \Longrightarrow L$$

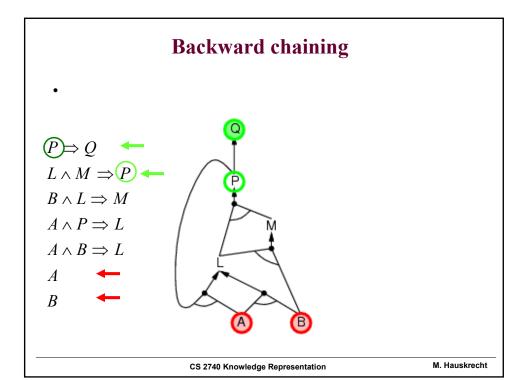
$$A \wedge B \Rightarrow L$$

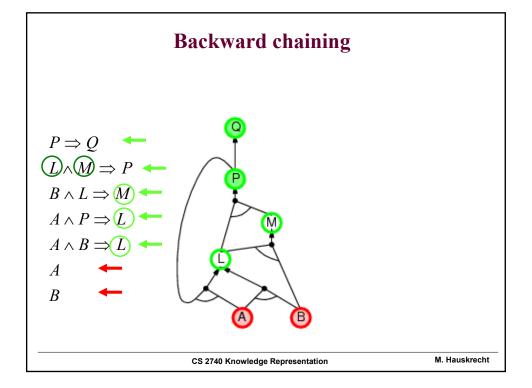
$$A \longleftarrow$$

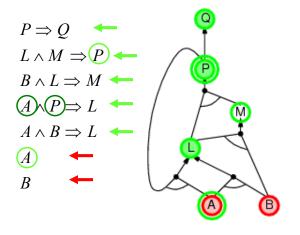
$$B \longrightarrow$$



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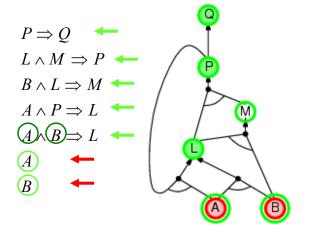




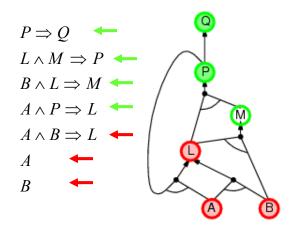
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Backward chaining



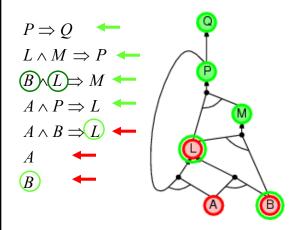
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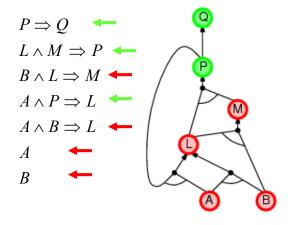
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Backward chaining



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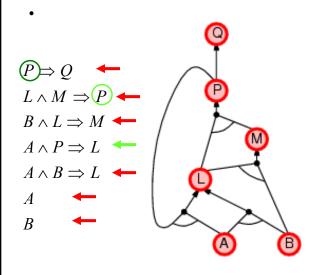
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Backward chaining

 $P \Rightarrow Q$ $D \land M \Rightarrow P$ $B \land L \Rightarrow M$ $A \land P \Rightarrow L$ $A \land B \Rightarrow L$ A B

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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,
 - e.g., Where are my keys? How do I get into a PhD program?
- Complexity of BC can be much less than linear in size of KB

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KB agents based on propositional logic

- Propositional logic allows us to build knowledge-based agents capable of answering queries about the world by infering new facts from the known ones
- **Example:** an agent for diagnosis of a bacterial disease

Facts: The stain of the organism is gram-positive

The growth conformation of the organism is chains

Rules: (If) The stain of the organism is gram-positive \land

The morphology of the organism is coccus \land

The growth conformation of the organism is chains

(Then) \Rightarrow The identity of the organism is streptococcus

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Example. Animal identification system.

- I1. If the animal has hair then it is a mammal
- I2. If the animal gives milk then it is a mammal
- I3. If the animal has feathers then it is a bird
- I4. If the animal flies and it lays eggs then it is a bird
- 15. If the animal is a mammal and it eats meat then it is a carnivore
- 16. If the animal is a mammal and it has pointed teeth and it has claws and its eyes point forward then it is a carnivore
- 17. If the animal is a mammal and it has hoofs then it is an ungulate
- 18. If the animal is a mammal and it chews cud then it is an ungulate and it is even-toed
- 19. If the animal is a carnivore and it has a tawny color and it has dark spots then it is a cheetah
- 110. If the animal is a carnivore and it has a tawny color and it has black strips then it is a tiger
- III. If the animal is an ungulate and it has long legs and it has a long neck and it has a tawny color and it has dark spots then it is a giraffe
- I12. If the animal is an ungulate and it has a white color and it has black stripes then it is a zebra
- II3. If the animal is a bird and it does not fly and it has long legs and it has a long neck and it is black and white then it is an ostrich,
- II4. If the animal is a bird and it does not fly and it swims and it is black and white then it is a penguin
- II5. If the animal is a bird and it is a good flyer then it is an albatross.