

# Propositional Logic: Methods of Proof (Part II)

# You will be expected to know

- Basic definitions
  - Inference, derive, sound, complete
- Conjunctive Normal Form (CNF)
  - Convert a Boolean formula to CNF
- Do a short resolution proof
- Horn Clauses
- ~~Do a short forward-chaining proof - - - - -~~
- ~~Do a short backward-chaining proof - - - - -~~
- ~~Model checking with backtracking search~~
- ~~Model checking with local search~~

# Review: Inference in Formal Symbol Systems

## Ontology, Representation, Inference

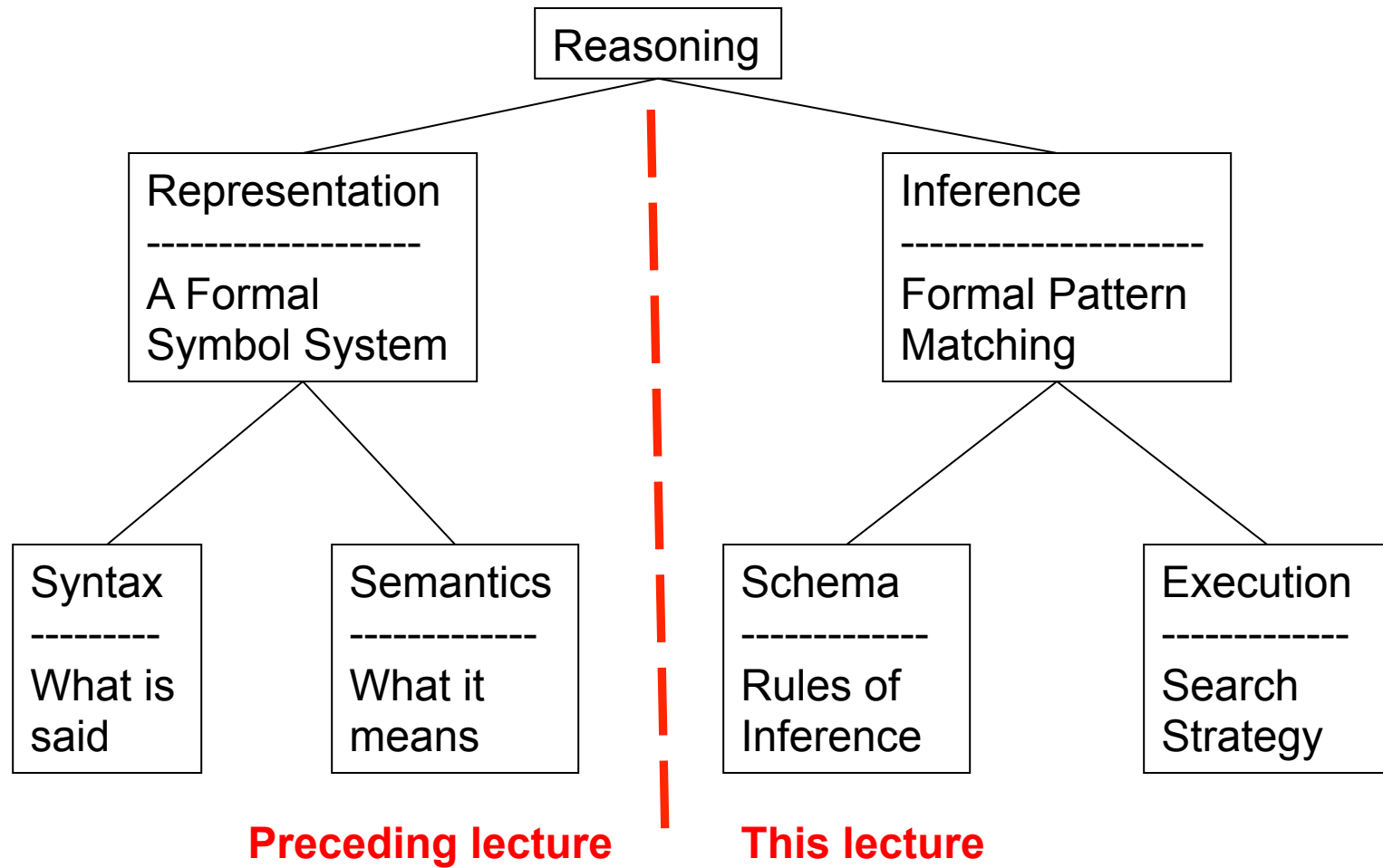
- **Formal Symbol Systems**
  - **Symbols** correspond to **things/ideas** in the world
  - **Pattern matching & rewrite** corresponds to **inference**
- **Ontology:** What exists in the world?
  - What must be represented?
- **Representation:** Syntax vs. Semantics
  - What's Said vs. What's Meant
- **Inference:** Schema vs. Mechanism
  - Proof Steps vs. Search Strategy

## Ontology:

What kind of things exist in the world?

What do we need to describe and reason about?

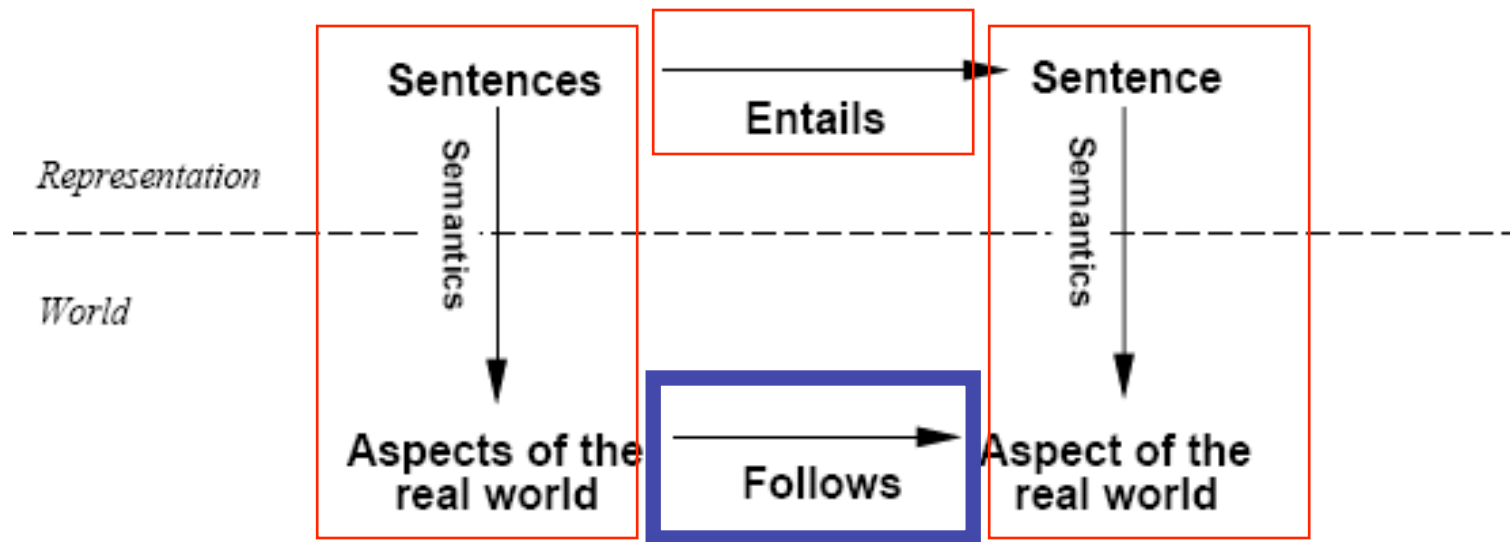
# Review



# Review

- Definitions:
  - Syntax, Semantics, Sentences, Propositions, Entails, Follows, Derives, Inference, Sound, Complete, Model, Satisfiable, Valid (or Tautology)
- Syntactic Transformations:
  - E.g.,  $(A \Rightarrow B) \Leftrightarrow (\neg A \vee B)$
- Semantic Transformations:
  - E.g.,  $(KB \models \alpha) \equiv (\models (KB \Rightarrow \alpha))$
- Truth Tables
  - Negation, Conjunction, Disjunction, Implication, Equivalence (Biconditional)
  - Inference by Model Enumeration

# Review: Schematic perspective



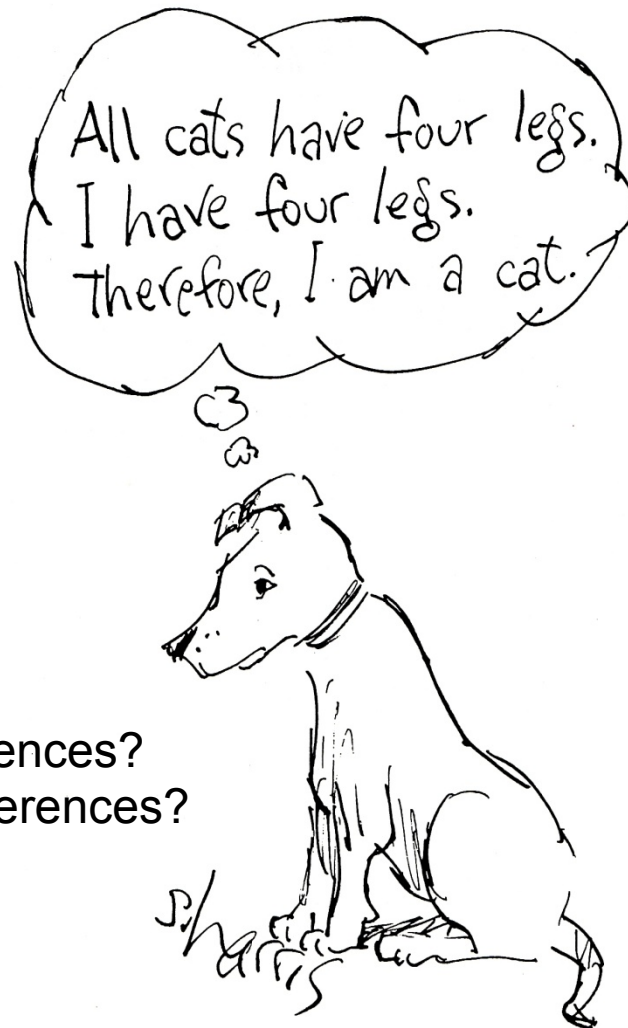
*If KB is true in the real world,  
then any sentence  $\alpha$  **entailed** by KB  
is also true in the real world.*

# So --- how do we keep it from “Just making things up.” ?

Is this inference correct?

How do you know?

How can you tell?



How can we **make correct** inferences?  
How can we **avoid incorrect** inferences?

“Einstein Simplified:  
Cartoons on Science”  
by Sydney Harris, 1992,  
Rutgers University Press

# So --- how do we keep it from “Just making things up.” ?

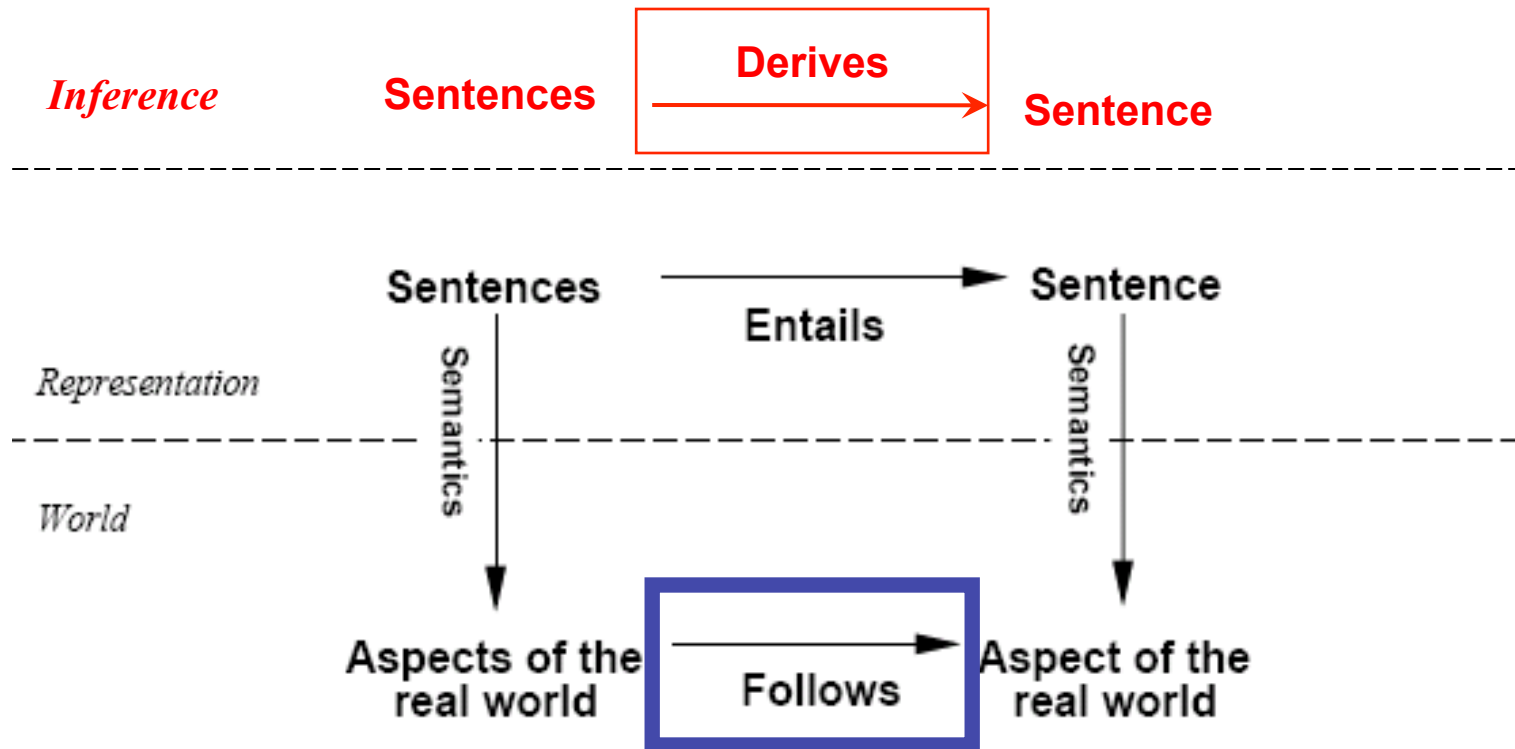
Is this inference correct?

- All men are people;  
Half of all people are women;  
Therefore, half of all men are women.  

How do you know?  
How can you tell?
- Penguins are black and white;  
Some old TV shows are black and white;  
Therefore, some penguins are old TV shows.



# Schematic perspective



*If KB is true in the real world,  
then any sentence  $\alpha$  **derived** from KB  
**by a sound inference procedure**  
is also true in the real world.*

# Logical inference

- The notion of entailment can be used for logic inference.
  - Model checking (see wumpus example):  
enumerate all possible models and check whether  $\alpha$  is true.
- Sound (or *truth preserving*):  
The algorithm **only** derives entailed sentences.
  - Otherwise it just makes things up.  
 *$i$  is sound iff whenever  $KB \models_i \alpha$  it is also true that  $KB \models \alpha$*
  - *E.g., model-checking is sound*Refusing to infer any sentence is Sound; so, Sound is weak alone.
- Complete:  
The algorithm can derive **every** entailed sentence.  
 *$i$  is complete iff whenever  $KB \models \alpha$  it is also true that  $KB \models_i \alpha$*   
Deriving every sentence is Complete; so, Complete is weak alone.

# Proof methods

- Proof methods divide into (roughly) two kinds:

Application of inference rules:

Legitimate (sound) generation of new sentences from old.

- Resolution --- KB is in Conjunctive Normal Form (CNF)
- ~~Forward & Backward chaining~~ –

Model checking

Searching through truth assignments.

- ~~Improved backtracking: Davis–Putnam–Logemann–Loveland (DPLL)~~
- ~~Heuristic search in model space: Walksat.~~

# Examples of Sound Inference Patterns

## Classical Syllogism (due to Aristotle)

All Ps are Qs  
X is a P  
Therefore, X is a Q

All Men are Mortal  
Socrates is a Man  
Therefore, Socrates is Mortal

## Implication (Modus Ponens)

P implies Q  
P  
Therefore, Q

Smoke implies Fire  
Smoke  
Therefore, Fire

Why is this different from:  
All men are people  
Half of people are women  
So half of men are women

## Contrapositive (Modus Tollens)

P implies Q  
Not Q  
Therefore, Not P

Smoke implies Fire  
Not Fire  
Therefore, not Smoke

## Law of the Excluded Middle (due to Aristotle)

A Or B  
Not A  
Therefore, B

Alice is a Democrat or a Republican  
Alice is not a Democrat  
Therefore, Alice is a Republican

# Inference by Resolution

- KB is represented in CNF
  - KB = AND of all the sentences in KB
  - KB sentence = clause = OR of literals
  - Literal = propositional symbol or its negation
- Find two clauses in KB, one of which contains a literal and the other its negation
- Cancel the literal and its negation
- Bundle everything else into a new clause
- Add the new clause to KB

# Conjunctive Normal Form (CNF)

- Boolean formulae are central to CS
  - Boolean logic is the way our discipline works
- Two canonical Boolean formulae representations:
  - CNF = Conjunctive Normal Form
    - A conjunct of disjuncts = (AND (OR ...) (OR ...))
    - “...” = a list of literals (= a variable or its negation)
    - CNF is used by Resolution Theorem Proving
  - DNF = Disjunctive Normal Form
    - A disjunct of conjuncts = (OR (AND ...) (AND ...))
    - DNF is used by Decision Trees in Machine Learning
- Can convert any Boolean formula to CNF or DNF

Clause



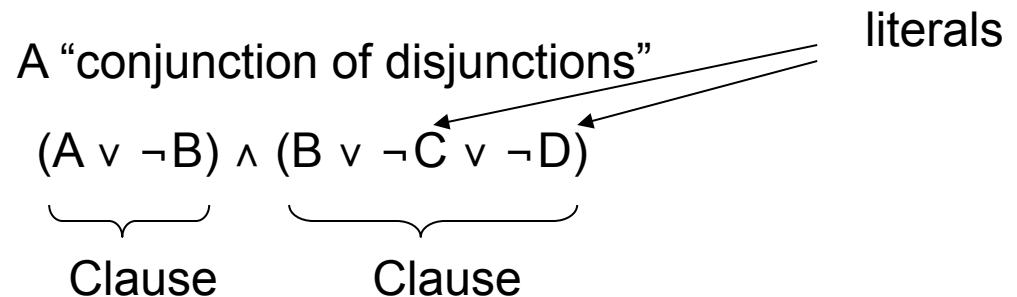
Term



# Conjunctive Normal Form (CNF)

We'd like to prove:  $KB \models \alpha$   
(This is equivalent to  $KB \wedge \neg \alpha$  is unsatisfiable.)

We first rewrite  $KB \wedge \neg \alpha$  into **conjunctive normal form (CNF)**.



- Any KB can be converted into CNF.
- In fact, any KB can be converted into CNF-3 using clauses with at most 3 literals.

# Example: Conversion to CNF

Example:  $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$

1. Eliminate  $\Leftrightarrow$  by replacing  $\alpha \Leftrightarrow \beta$  with  $(\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)$ .  
 $= (B_{1,1} \Rightarrow (P_{1,2} \vee P_{2,1})) \wedge ((P_{1,2} \vee P_{2,1}) \Rightarrow B_{1,1})$
2. Eliminate  $\Rightarrow$  by replacing  $\alpha \Rightarrow \beta$  with  $\neg\alpha \vee \beta$  and simplify.  
 $= (\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg(P_{1,2} \vee P_{2,1}) \vee B_{1,1})$
3. Move  $\neg$  inwards using de Morgan's rules and simplify.  
$$\neg(\alpha \vee \beta) = \neg\alpha \wedge \neg\beta$$
$$= (\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge ((\neg P_{1,2} \wedge \neg P_{2,1}) \vee B_{1,1})$$
4. Apply distributive law ( $\wedge$  over  $\vee$ ) and simplify.  
 $= (\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg P_{1,2} \vee B_{1,1}) \wedge (\neg P_{2,1} \vee B_{1,1})$



# Example: Conversion to CNF

Example:  $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$

From the previous slide we had:

$$= (\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg P_{1,2} \vee B_{1,1}) \wedge (\neg P_{2,1} \vee B_{1,1})$$

5. KB is the conjunction of all of its sentences (all are true),  
so write each clause (disjunct) as a sentence in KB:

KB =

...

$$(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1})$$

$$(\neg P_{1,2} \vee B_{1,1})$$

$$(\neg P_{2,1} \vee B_{1,1})$$

...



(same)

Often, Won't Write "v" or "&"  
(we know they are there)

$$\left( \begin{array}{l} \neg B_{1,1} \quad P_{1,2} \\ \neg P_{1,2} \quad B_{1,1} \\ \neg P_{2,1} \quad B_{1,1} \end{array} \right) P_{2,1}$$

# Inference by Resolution

- KB is represented in CNF
  - KB = AND of all the sentences in KB
  - KB sentence = clause = OR of literals
  - Literal = propositional symbol or its negation
- Find two clauses in KB, one of which contains a literal and the other its negation
- Cancel the literal and its negation
- Bundle everything else into a new clause
- Add the new clause to KB

# Resolution = Efficient Implication

Recall that  $(A \Rightarrow B) = ((\text{NOT } A) \text{ OR } B)$

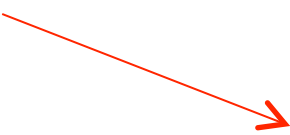
and so:

$$\begin{aligned} (Y \text{ OR } X) &= ((\text{NOT } X) \Rightarrow Y) \\ ((\text{NOT } Y) \text{ OR } Z) &= (Y \Rightarrow Z) \end{aligned}$$

which yields:

$$((Y \text{ OR } X) \text{ AND } ((\text{NOT } Y) \text{ OR } Z)) = ((\text{NOT } X) \Rightarrow Z) = (X \text{ OR } Z)$$

$(\text{OR } A \text{ B C D})$	$\rightarrow \text{Same} \rightarrow$	$(\text{NOT } (\text{OR } B \text{ C D})) \Rightarrow A$
$(\text{OR } \neg A \text{ E F G})$	$\rightarrow \text{Same} \rightarrow$	$A \Rightarrow (\text{OR } E \text{ F G})$

$(\text{OR } B \text{ C D E F G})$		$(\text{NOT } (\text{OR } B \text{ C D})) \Rightarrow (\text{OR } E \text{ F G})$
		$(\text{OR } B \text{ C D E F G})$

Recall: All clauses in KB are conjoined by an implicit AND (= CNF representation).

# Resolution Examples

- **Resolution:** inference rule for CNF: **sound and complete!** \*

$(A \vee B \vee C)$

$(\neg A)$

“If A or B or C is true, but not A, then B or C must be true.”

-----

$\therefore (B \vee C)$

$(A \vee B \vee C)$

$(\neg A \vee D \vee E)$

“If A is false then B or C must be true, or if A is true then D or E must be true, hence since A is either true or false, B or C or D or E must be true.”

-----

$\therefore (B \vee C \vee D \vee E)$

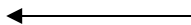
$(A \vee B)$

$(\neg A \vee B)$

“If A or B is true, and not A or B is true, then B must be true.”

-----

$\therefore (B \vee B) \equiv B$



Simplification  
is done always.

\* Resolution is “refutation complete” in that it can prove the truth of any entailed sentence by refutation.

\* You can start two resolution proofs in parallel, one for the sentence and one for its negation, and see which branch returns a correct proof.

# Only Resolve ONE Literal Pair!

If more than one pair, result always = TRUE.

**Useless!!** Always simplifies to TRUE!!

**No!**

(OR A B C D)  
(OR  $\neg A$   $\neg B$  F G)

---

(OR C D F G)

**No!**

**No!**

(OR A B C D)  
(OR  $\neg A$   $\neg B$   $\neg C$  )

---

(OR D)

**No!**

**Yes! (but = TRUE)**

(OR A B C D)  
(OR  $\neg A$   $\neg B$  F G)

---

(OR B  $\neg B$  C D F G)

**Yes! (but = TRUE)**

**Yes! (but = TRUE)**

(OR A B C D)  
(OR  $\neg A$   $\neg B$   $\neg C$  )

---

(OR A  $\neg A$  B  $\neg B$  D)

**Yes! (but = TRUE)**

# Resolution Algorithm

- The resolution algorithm tries to prove:  $KB \models \alpha$  *equivalent to*  
 $KB \wedge \neg \alpha$  *unsatisfiable*
- Generate all new sentences from KB and the (negated) query.
- One of two things can happen:
  1. We find  $P \wedge \neg P$  which is unsatisfiable. I.e. we can entail the query.
  2. We find no contradiction: there is a model that satisfies the sentence  $KB \wedge \neg \alpha$  (non-trivial) and hence we cannot entail the query.

# Resolution example

Stated in English

- “Laws of Physics” in the Wumpus World:
  - “A breeze in B11 is equivalent to a pit in P12 or a pit in P21.”
- Particular facts about a specific instance:
  - “There is no breeze in B11.”
- Goal or query sentence:
  - “Is it true that P12 does not have a pit?”

# Resolution example

Stated in Propositional Logic

- “Laws of Physics” in the Wumpus World:
  - “A breeze in B11 is equivalent to a pit in P12 or a pit in P21.”

$$(B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1}))$$

We converted this sentence to CNF in the CNF example we worked above.

- Particular facts about a specific instance:
  - “There is no breeze in B11.”

$$(\neg B_{1,1})$$

- Goal or query sentence:
  - “Is it true that P12 does not have a pit?”

$$(\neg P_{1,2})$$



# Resolution example

Resulting Knowledge Base stated in CNF

- “Laws of Physics” in the Wumpus World:

$$\begin{pmatrix} \neg B_{1,1} & P_{1,2} & P_{2,1} \\ \neg P_{1,2} & B_{1,1} \\ \neg P_{2,1} & B_{1,1} \end{pmatrix}$$

- Particular facts about a specific instance:

$$(\neg B_{1,1})$$

- Negated goal or query sentence:

$$(P_{1,2})$$

# Resolution example

A Resolution proof ending in ( )

- Knowledge Base at start of proof:

$(\neg B_{1,1} \quad P_{1,2} \quad P_{2,1})$

$(\neg P_{1,2} \quad B_{1,1})$

$(\neg P_{2,1} \quad B_{1,1})$

$(\neg B_{1,1})$

$(P_{1,2})$

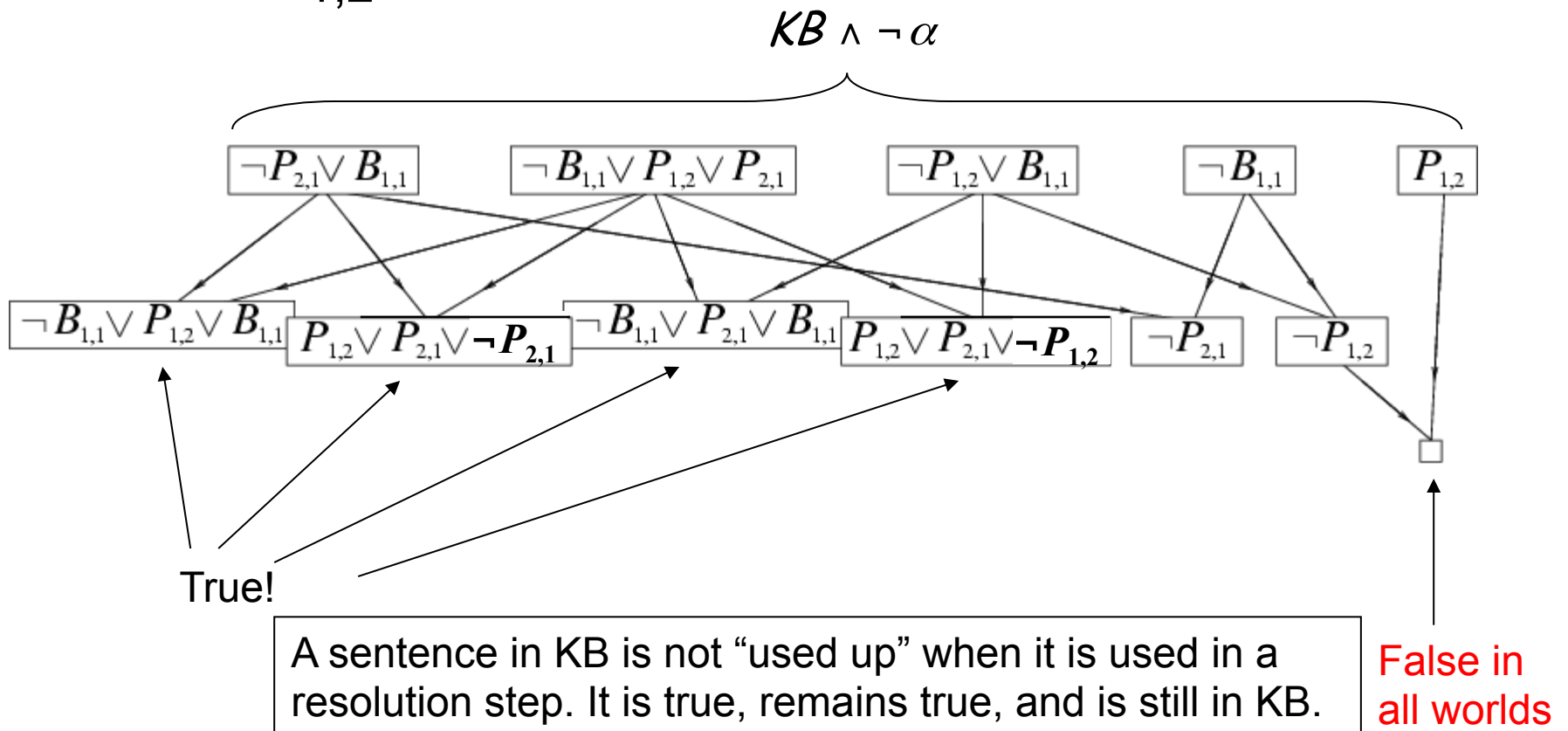
**A resolution proof ending in ( ):**

- Resolve  $(\neg P_{1,2} \quad B_{1,1})$  and  $(\neg B_{1,1})$  to give  $(\neg P_{1,2})$
- Resolve  $(\neg P_{1,2})$  and  $(P_{1,2})$  to give ( )
- Consequently, the goal or query sentence is entailed by KB.
- Of course, there are many other proofs, which are OK iff correct.

# Resolution example

Graphical view of the proof

- $KB = (B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})) \wedge \neg B_{1,1}$
- $\alpha = \neg P_{1,2}$



# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*  
*Prove that the unicorn is both magical and horned.*

Problem 7.2, R&N page 280. (Adapted from Barwise and Etchemendy, 1993.)

Note for non-native-English speakers: immortal = not mortal

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned. Prove that the unicorn is both magical and horned.*
- **First, Ontology:** What do we need to describe and reason about?
- Use these propositional variables (“immortal” = “not mortal”):
  - Y = unicorn is mYthical                      R = unicorn is moRtal
  - M = unicorn is a maMmal                      H = unicorn is Horned
  - G = unicorn is maGical

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that the unicorn is both magical and horned.

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmal

H = unicorn is Horned

G = unicorn is maGical

- **Second, translate to Propositional Logic, then to CNF:**
- Propositional logic (prefix form, aka Polish notation):
  - **(=> Y (NOT R) )** ; same as ( Y => (NOT R) ) in infix form
- CNF (clausal form) ; recall (A => B) = ( (NOT A) OR B )
  - **( (NOT Y) (NOT R) )**

Prefix form is often a better representation for a parser, since it looks at the first element of the list and dispatches to a handler for that operator token.

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that the unicorn is both magical and horned.

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmamal

H = unicorn is Horned

G = unicorn is maGical

- **Second, translate to Propositional Logic, then to CNF:**

- Propositional logic (prefix form):

– **( $\Rightarrow$  (NOT Y) (AND R M) )** ; same as ( (NOT Y)  $\Rightarrow$  (R AND M) ) in infix form

- CNF (clausal form)

– **(M Y)**

– **(R Y)**

If you ever have to do this “for real” you will likely invent a new domain language that allows you to state important properties of the domain --- then parse that into propositional logic, and then CNF.

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that the unicorn is both magical and horned.

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmal

H = unicorn is Horned

G = unicorn is maGical

- **Second, translate to Propositional Logic, then to CNF:**
- Propositional logic (prefix form):
  - **$(\Rightarrow (\text{OR } (\text{NOT } R) M) H)$**  ; same as  $((\text{Not } R) \text{ OR } M) \Rightarrow H$  in infix form
- CNF (clausal form)
  - **$(H (\text{NOT } M) )$**
  - **$(H R)$**



# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that the unicorn is both magical and horned.

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmal

H = unicorn is Horned

G = unicorn is maGical

- **Second, translate to Propositional Logic, then to CNF:**
- Propositional logic (prefix form)
  - **( $\Rightarrow$  H G)** ; same as  $H \Rightarrow G$  in infix form
- CNF (clausal form)
  - **( (NOT H) G )**

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that the unicorn is both magical and horned.

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmal

H = unicorn is Horned

G = unicorn is maGical

- **Current KB** (in CNF clausal form) =

( (NOT Y) (NOT R) )  
(H R)

(M Y)  
( (NOT H) G)

(R Y)

(H (NOT M) )

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that *the unicorn is both magical and horned.*

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmal

H = unicorn is Horned

G = unicorn is maGical

- **Third, negated goal to Propositional Logic, then to CNF:**
- Goal sentence in propositional logic (prefix form)
  - **(AND H G)** ; same as H AND G in infix form
- Negated goal sentence in propositional logic (prefix form)
  - **(NOT (AND H G) ) = (OR (NOT H) (NOT G) )**
- CNF (clausal form)
  - **( (NOT G) (NOT H) )**

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

*Prove that the unicorn is both magical and horned.*

Y = unicorn is mYthical

R = unicorn is moRtal

M = unicorn is a maMmal

H = unicorn is Horned

G = unicorn is maGical

- Current KB + negated goal (in CNF clausal form) =

( (NOT Y) (NOT R) )  
(H R)

(M Y)  
( (NOT H) G)

(R Y)                      (H (NOT M) )  
**( (NOT G) (NOT H) )**

# Detailed Resolution Proof Example

- **In words:** *If the unicorn is mythical, then it is immortal, but if it is not mythical, then it is a mortal mammal. If the unicorn is either immortal or a mammal, then it is horned. The unicorn is magical if it is horned.*

Prove that the unicorn is both magical and horned.

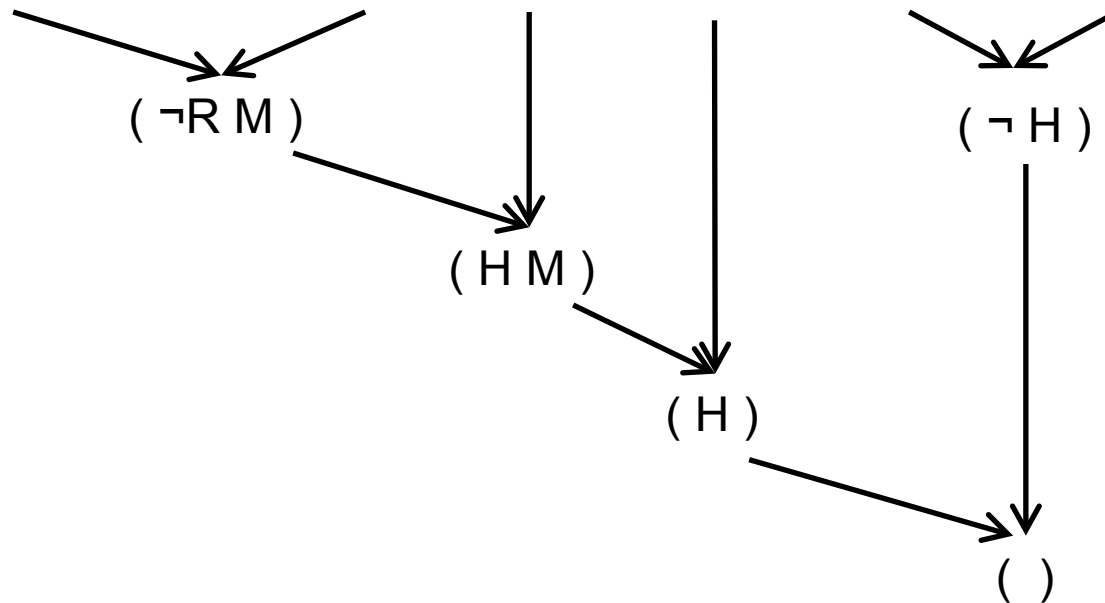
$(\neg Y) (\neg R)$	$(M Y)$	$(R Y)$	$(H (\neg M))$
$(H R)$	$(\neg H) G$	$(\neg G) (\neg H)$	

- **Fourth, produce a resolution proof ending in ( ):**
- Resolve  $(\neg H \neg G)$  and  $(\neg H G)$  to give  $(\neg H)$
- Resolve  $(\neg Y \neg R)$  and  $(Y M)$  to give  $(\neg R M)$
- Resolve  $(\neg R M)$  and  $(R H)$  to give  $(M H)$
- Resolve  $(M H)$  and  $(\neg M H)$  to give  $(H)$
- Resolve  $(\neg H)$  and  $(H)$  to give  $( )$
- Of course, there are many other proofs, which are OK iff correct.

# Detailed Resolution Proof Example

## Graph view of proof

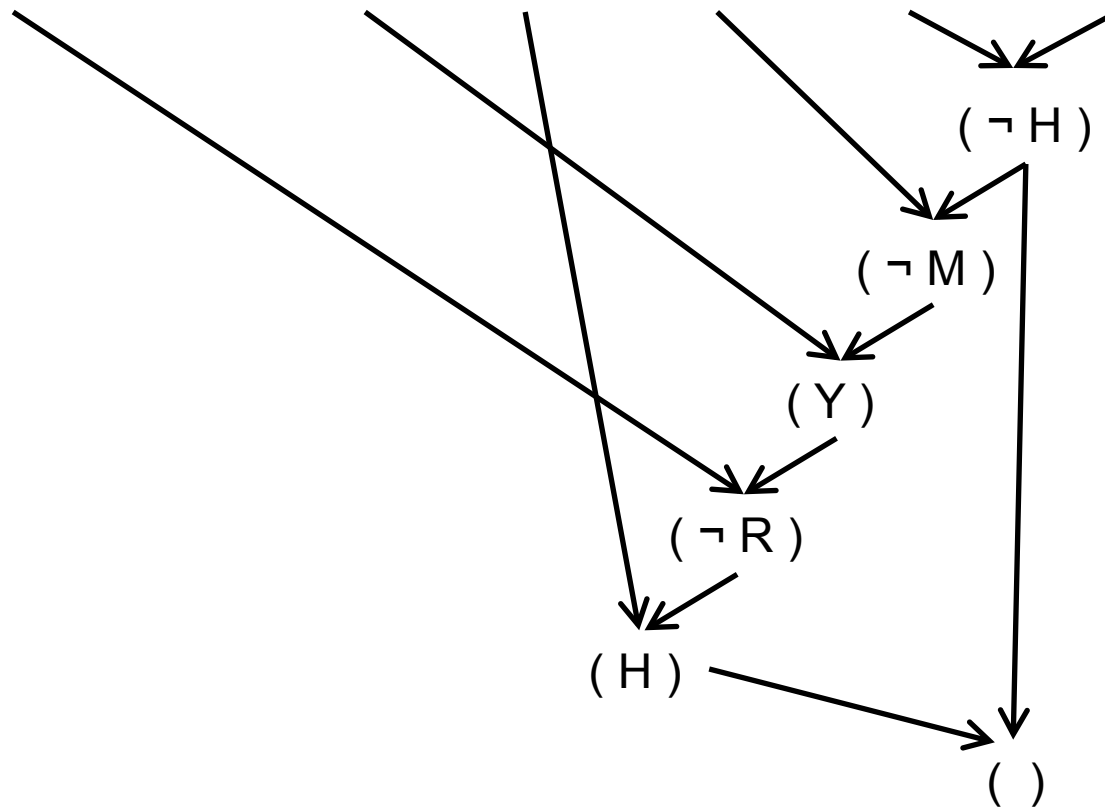
- $(\neg Y \neg R)(Y R)(Y M)(R H)(\neg M H)(\neg H G)(\neg G \neg H)$



# Detailed Resolution Proof Example

## Graph view of a different proof

- $(\neg Y \neg R)(Y R)(Y M)(R H)(\neg M H)(\neg H G)(\neg G \neg H)$



# Horn Clauses

- Resolution can be exponential in space and time.
- If we can reduce all clauses to “Horn clauses” inference is linear in space and time

A clause with at most 1 positive literal.

e.g.  $A \vee \neg B \vee \neg C$

- Every Horn clause can be rewritten as an implication with a conjunction of positive literals in the premises and at most a single positive literal as a conclusion.

e.g.  $A \vee \neg B \vee \neg C \equiv B \wedge C \Rightarrow A$

- 1 positive literal and  $\geq 1$  negative literal: definite clause (e.g., above)
- 0 positive literals: integrity constraint or goal clause

e.g.  $(\neg A \vee \neg B) \equiv (A \wedge B \Rightarrow \text{False})$  states that  $(A \wedge B)$  must be false

- 0 negative literals: fact

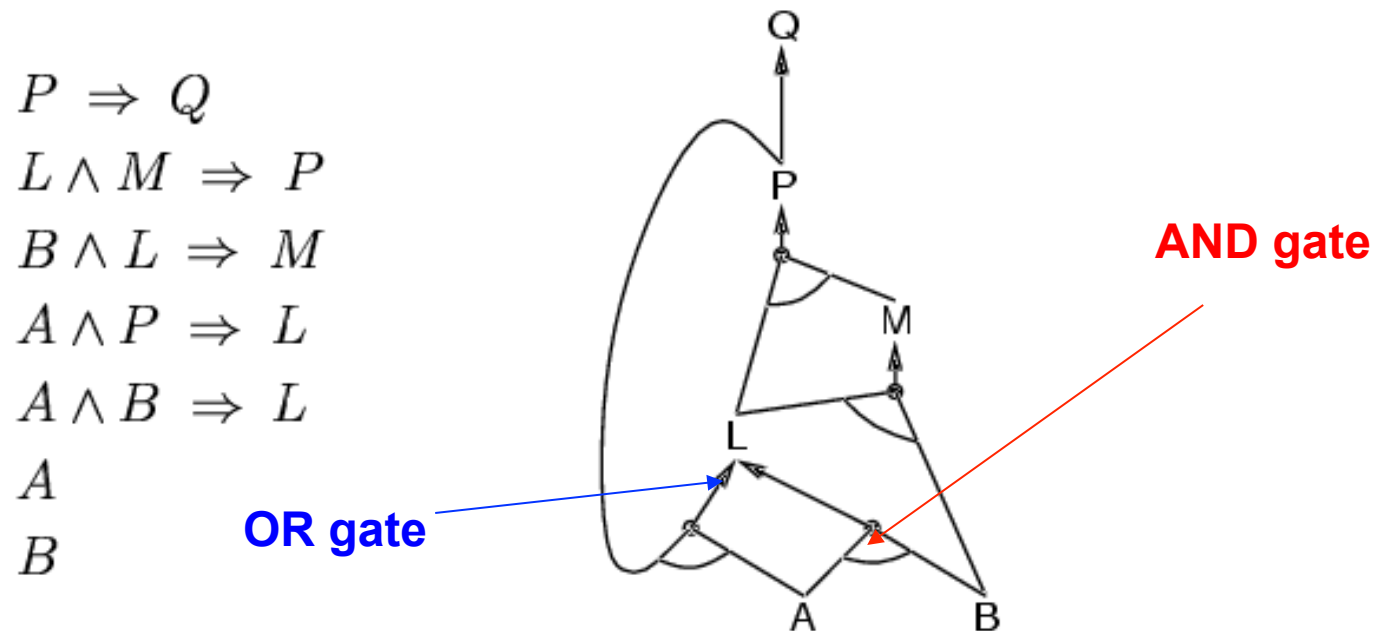
e.g.,  $(A) \equiv (\text{True} \Rightarrow A)$  states that A must be true.

- Forward Chaining and Backward chaining are sound and complete with Horn clauses and run linear in space and time.



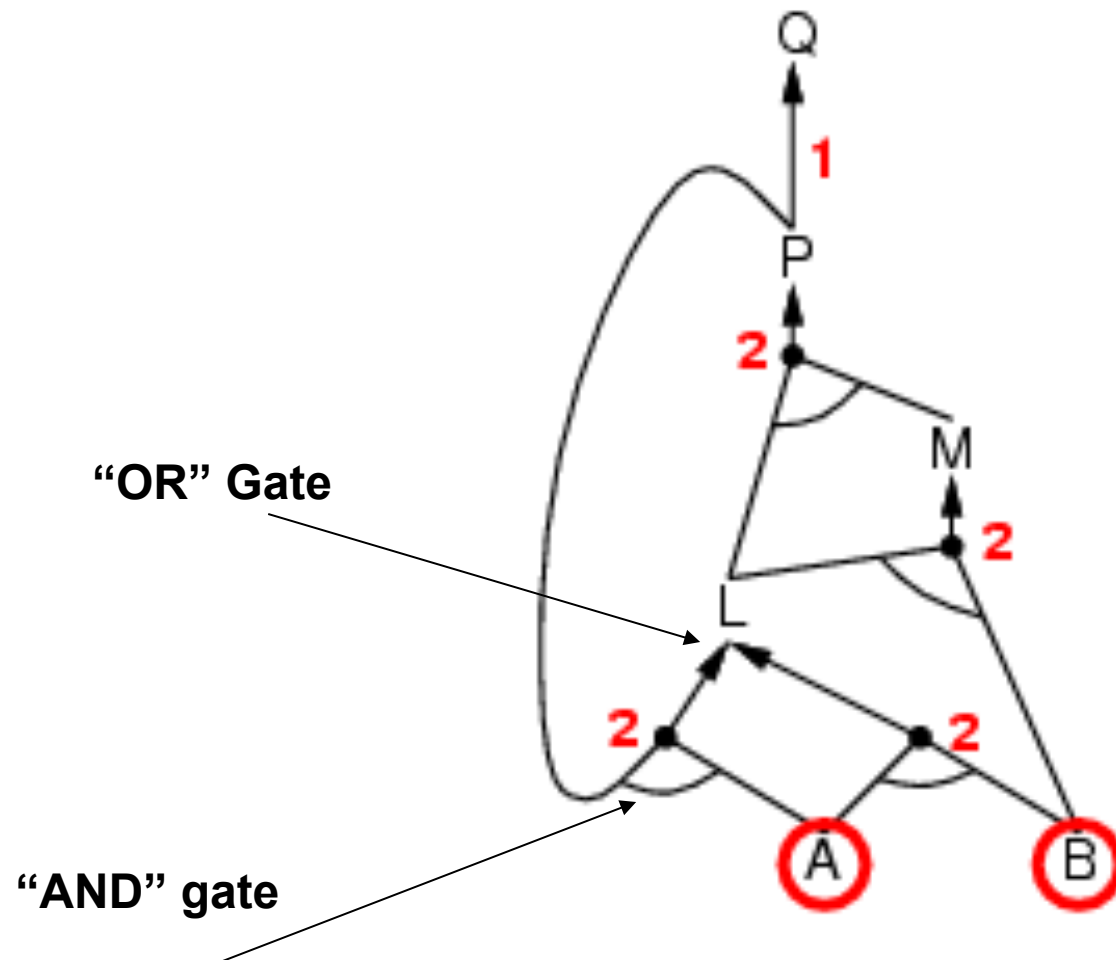
# Forward chaining (FC)

- Idea: fire any rule whose premises are satisfied in the *KB*, add its conclusion to the *KB*, until query is found.
- This proves that  $KB \Rightarrow Q$  is true in all possible worlds (i.e. trivial), and hence it proves entailment.

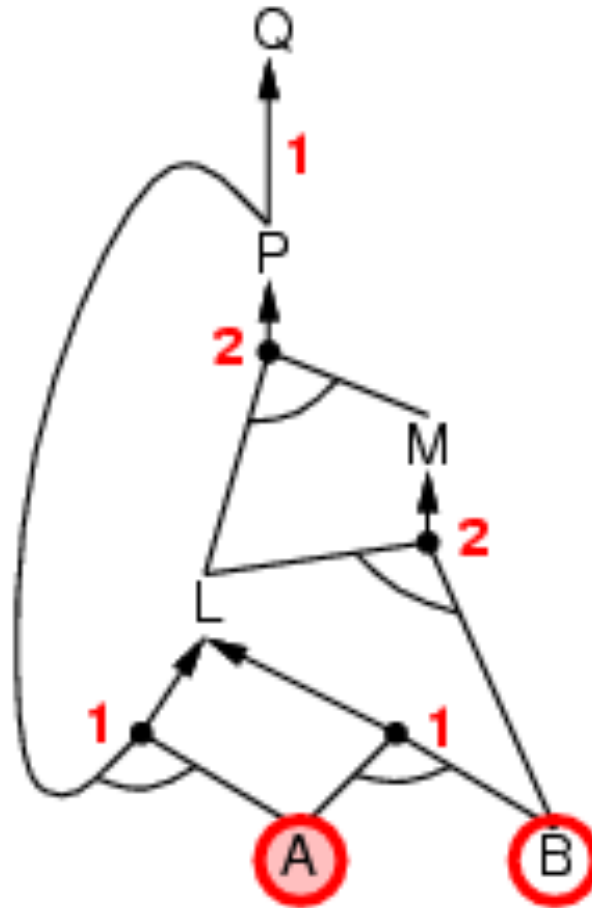


- Forward chaining is sound and complete for Horn KB

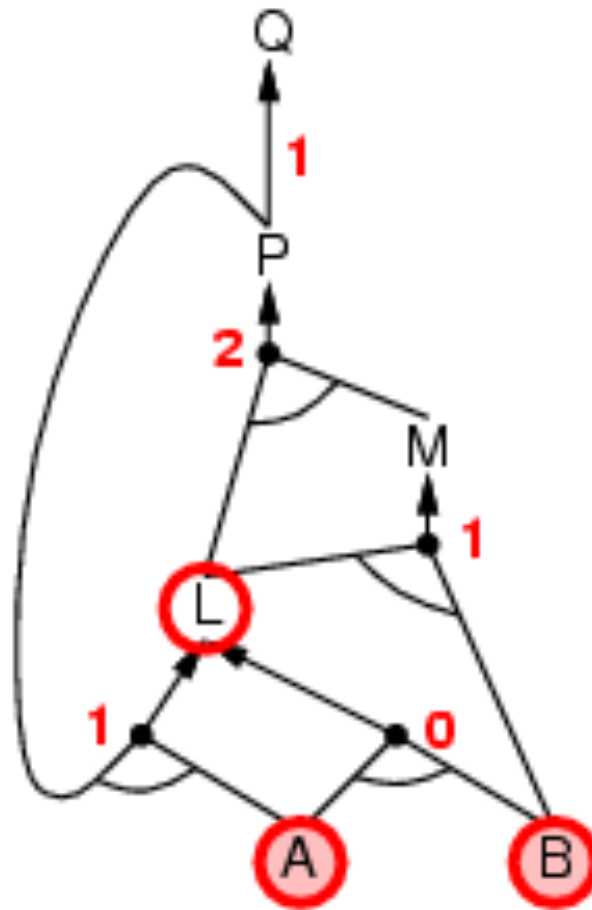
# Forward chaining example



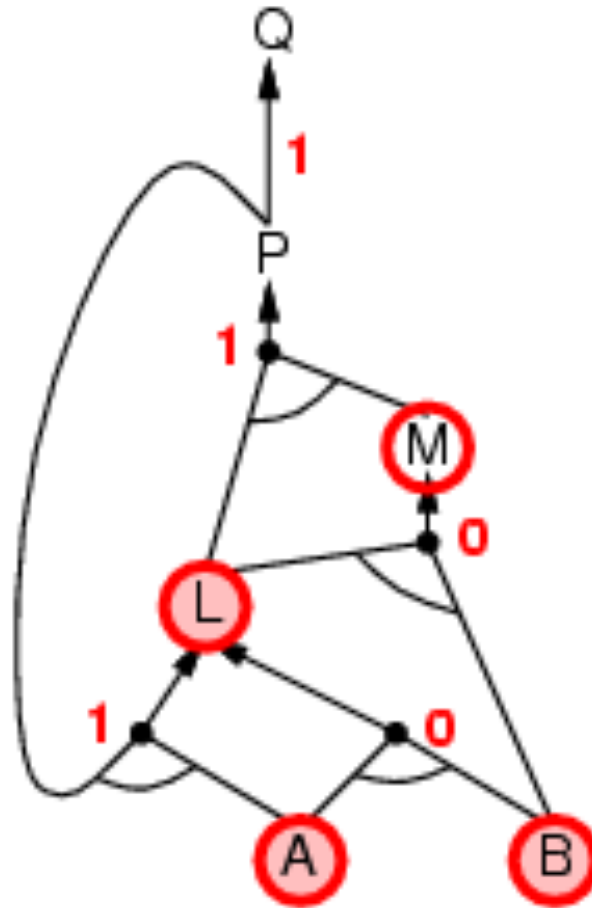
# Forward chaining example



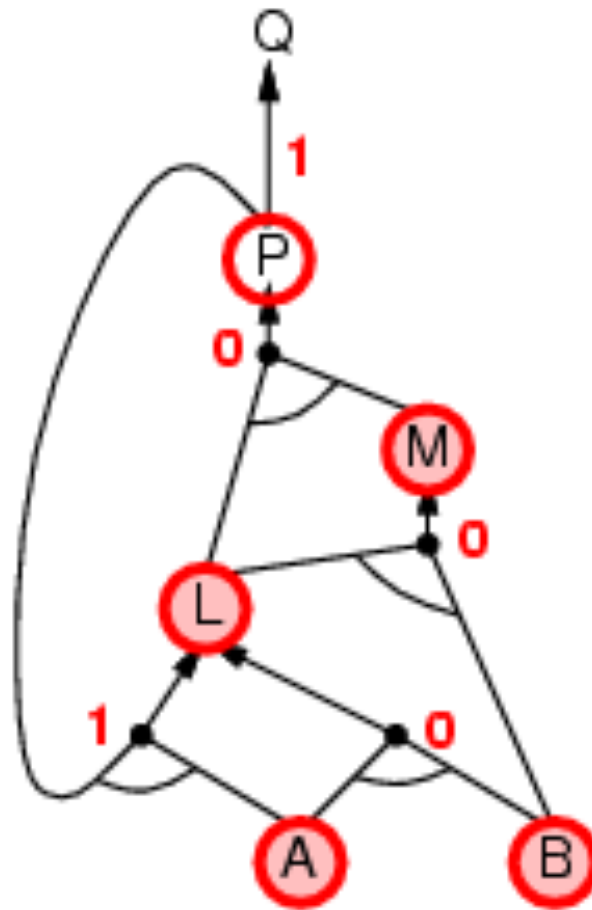
# Forward chaining example



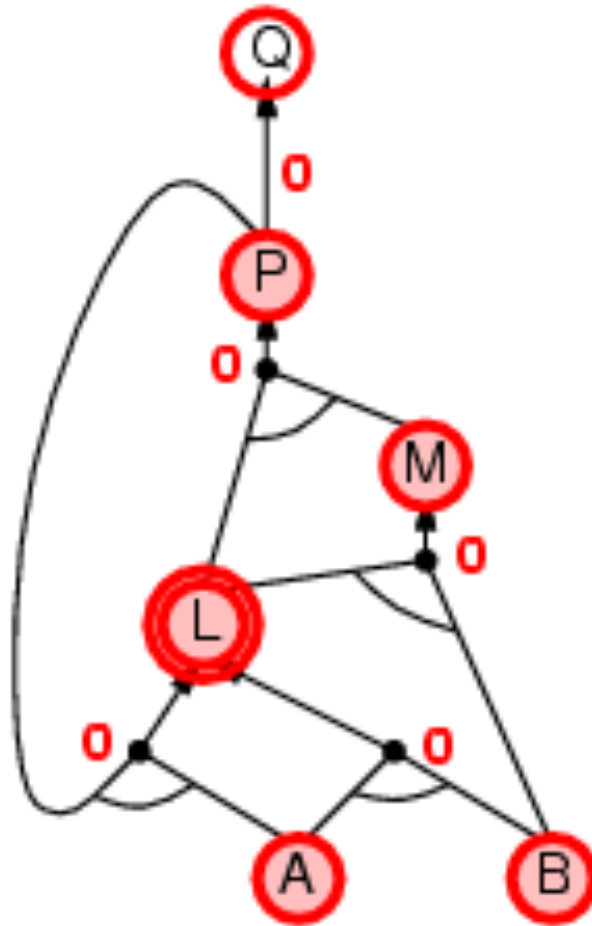
# Forward chaining example



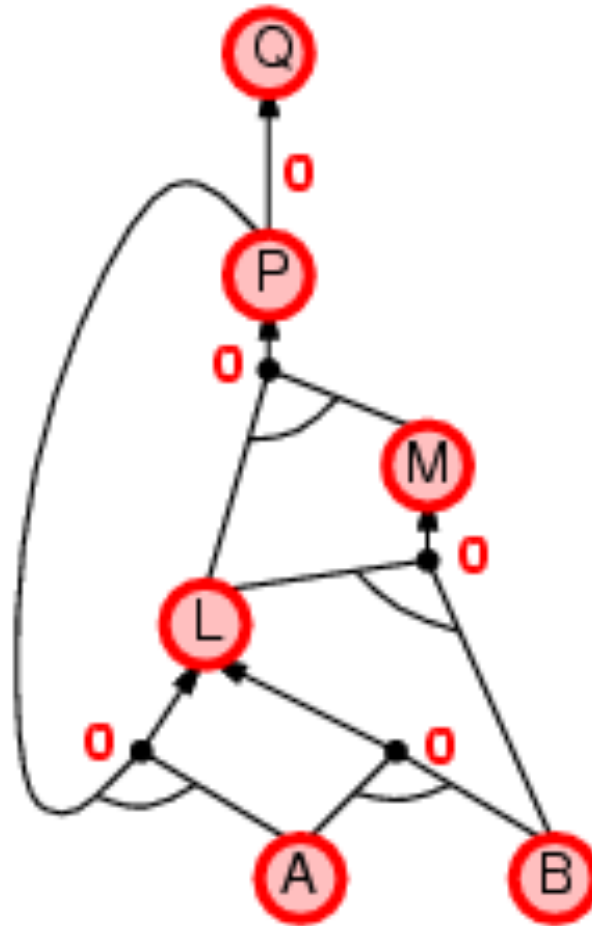
# Forward chaining example



# Forward chaining example



# Forward chaining example





# Backward chaining (BC)

Idea: work backwards from the query  $q$

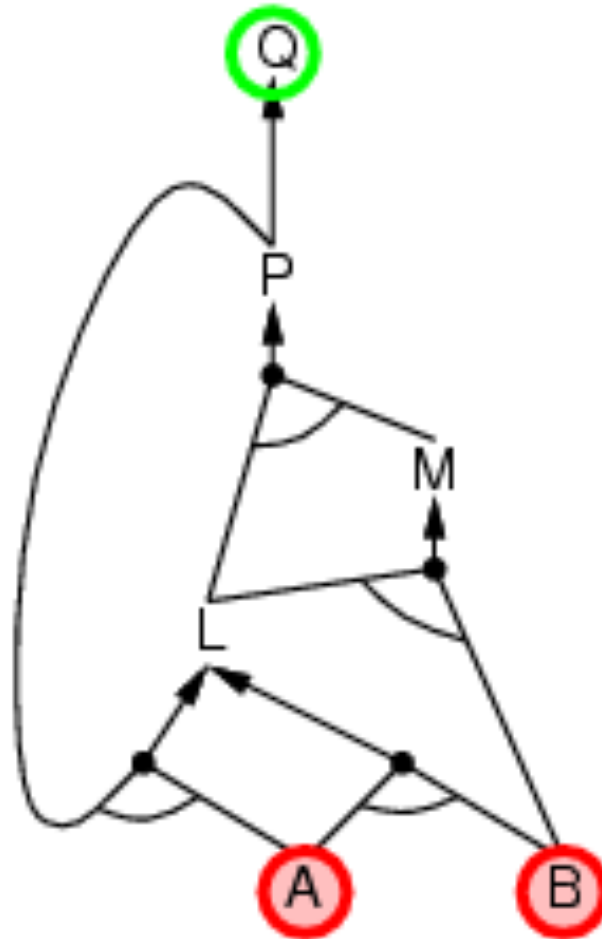
- check if  $q$  is known already, or
- prove by BC all premises of some rule concluding  $q$
- Hence BC maintains a stack of sub-goals that need to be proved to get to  $q$ .

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

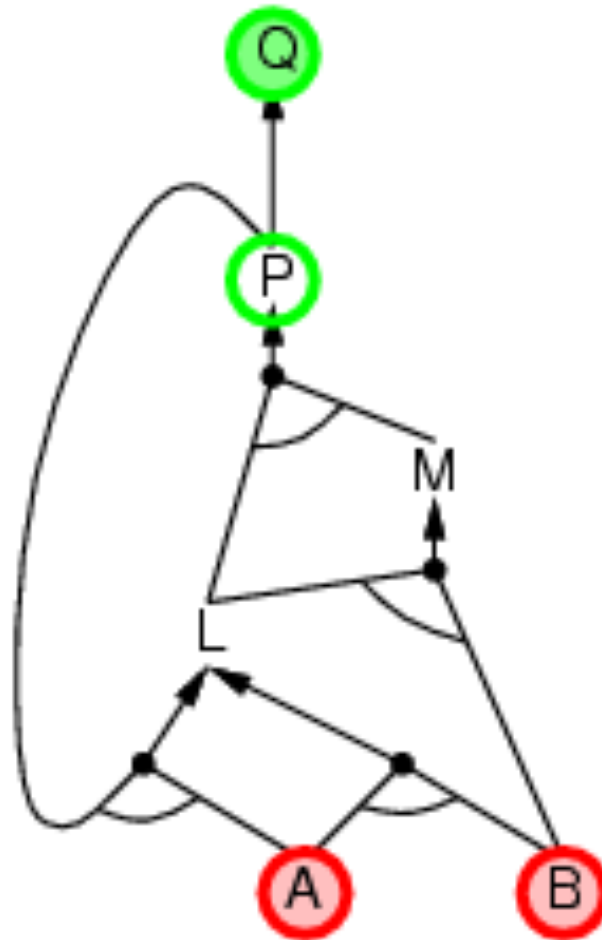
Avoid repeated work: check if new sub-goal

1. has already been proved true, or
2. has already failed

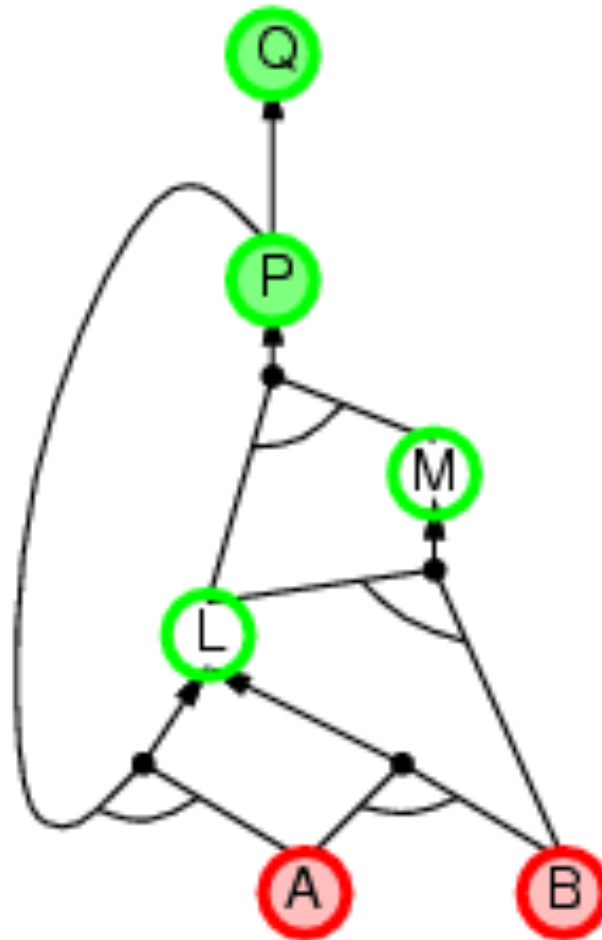
# Backward chaining example



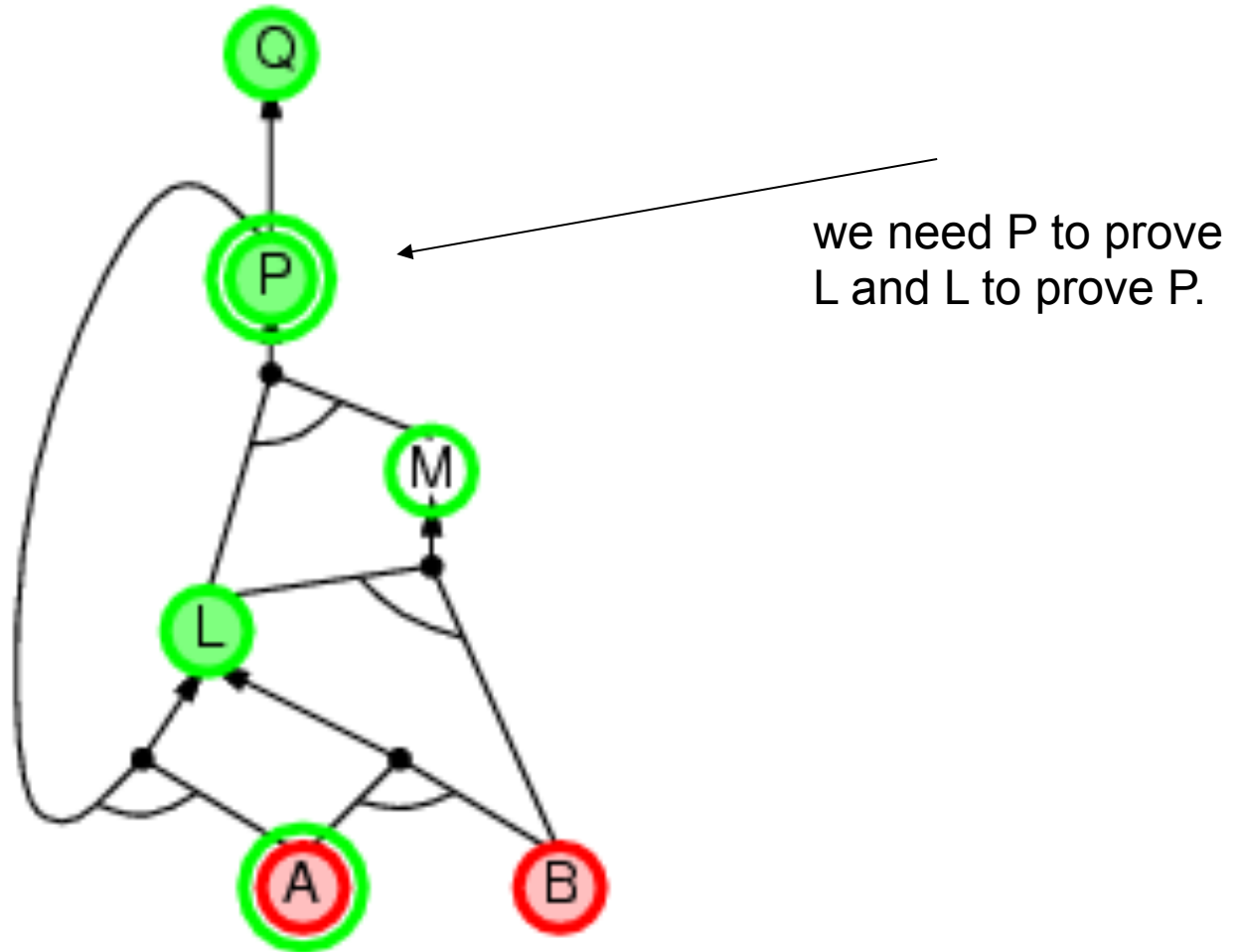
# Backward chaining example



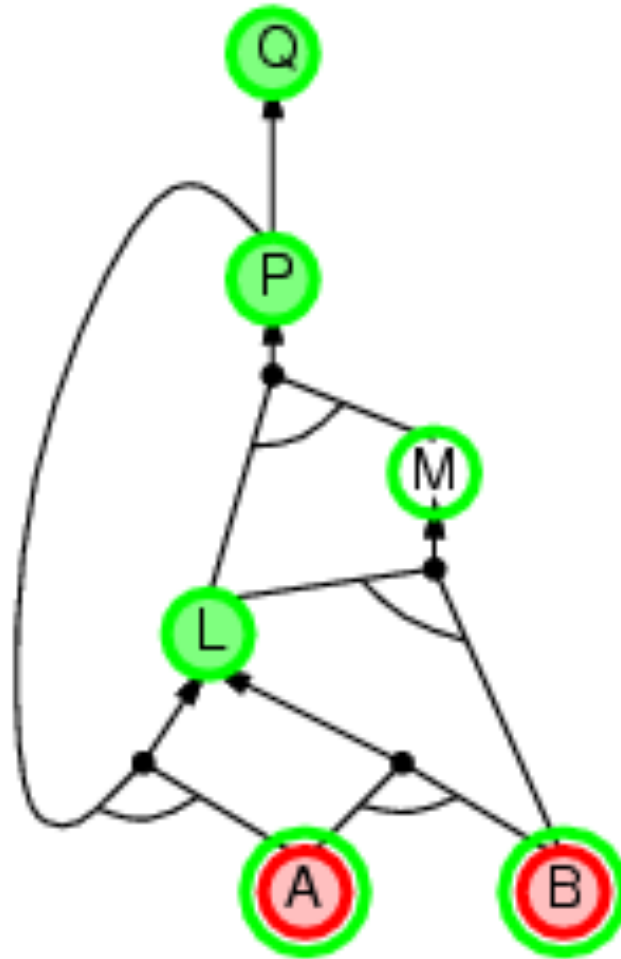
# Backward chaining example



# Backward chaining example

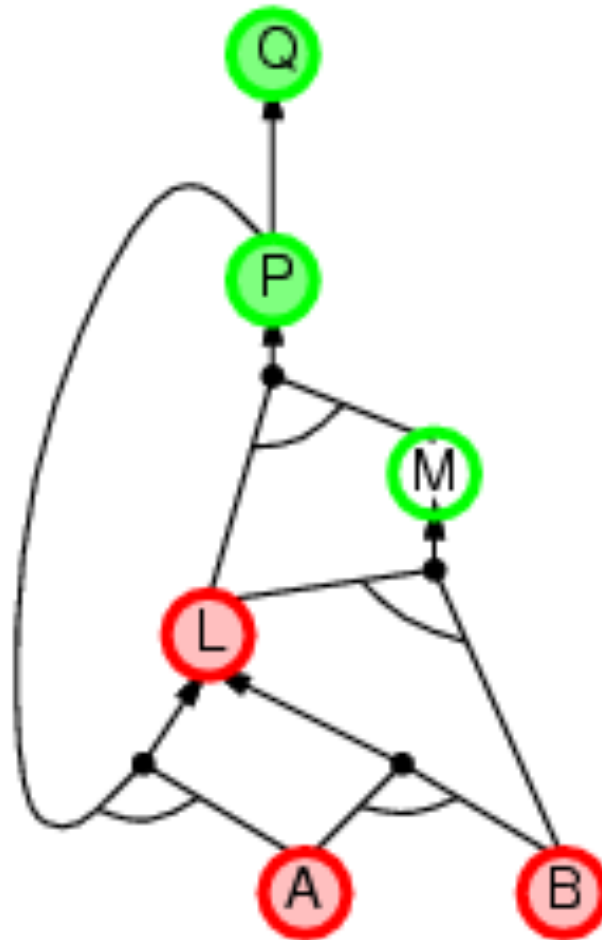


# Backward chaining example

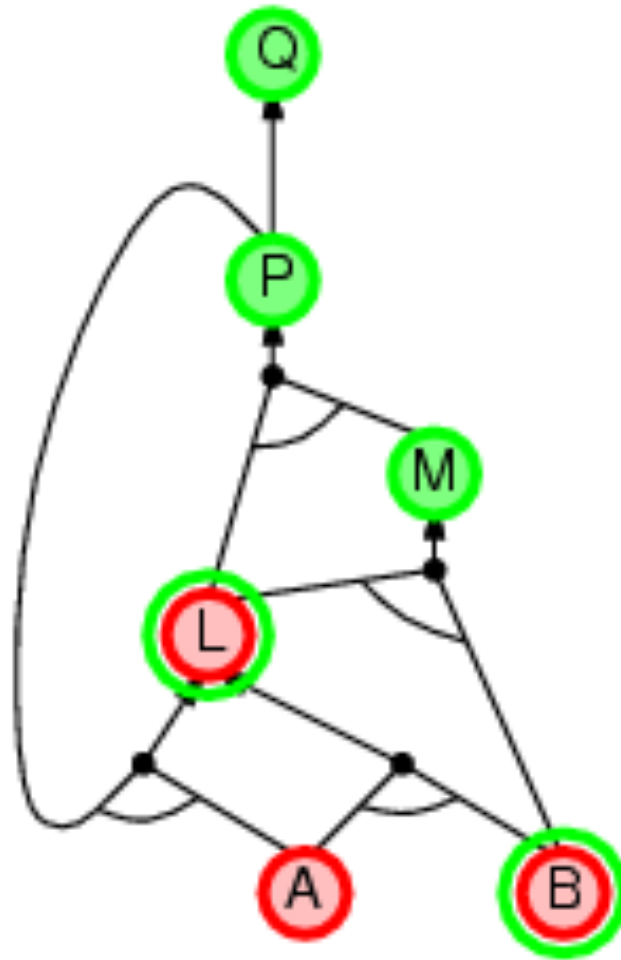


As soon as you can move forward, do so.

# Backward chaining example

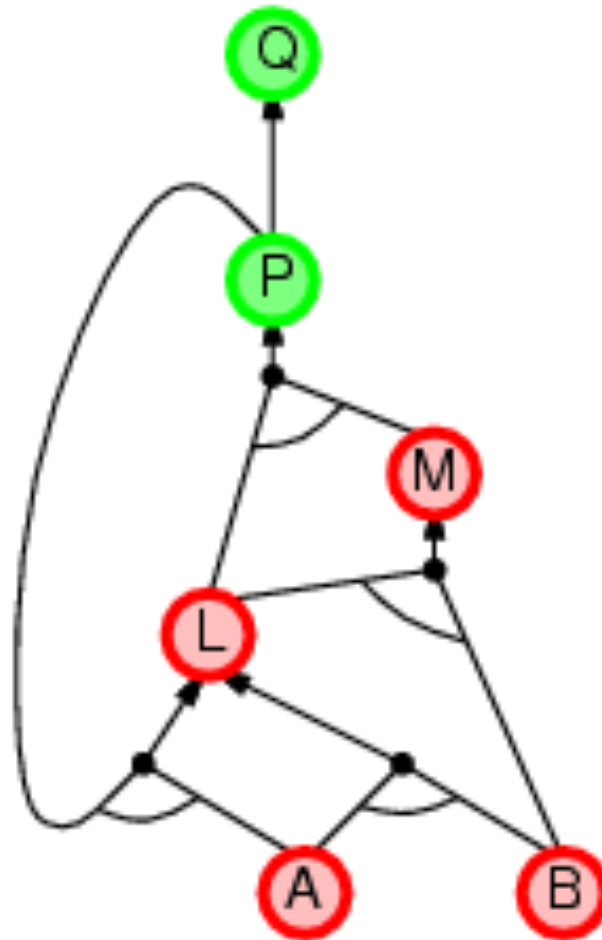


# Backward chaining example

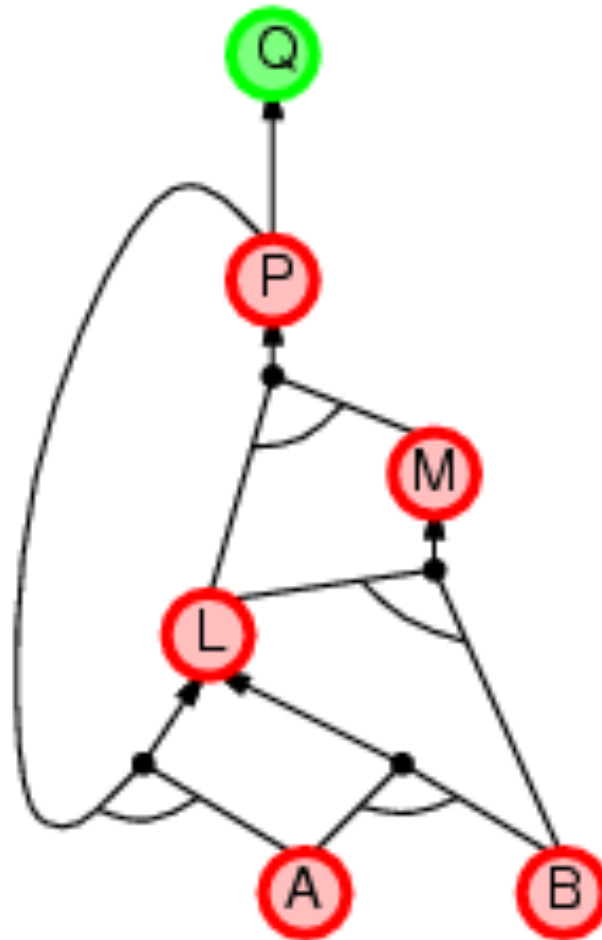




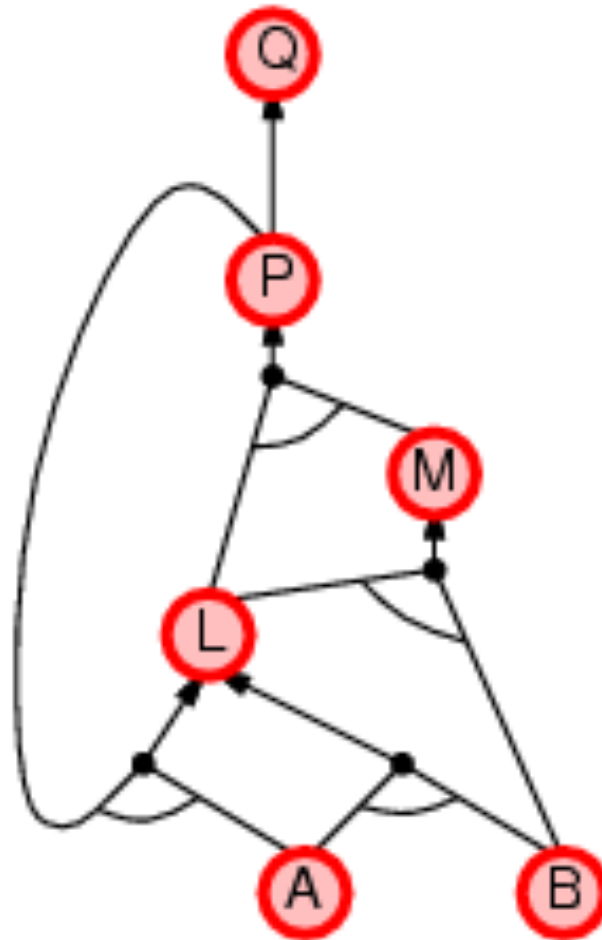
# Backward chaining example



# Backward chaining example



# Backward chaining example



# 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

# Model Checking

Two families of efficient algorithms:

- Complete backtracking search algorithms:
  - E.g., DPLL algorithm
- Incomplete local search algorithms
  - E.g., WalkSAT algorithm

# The DPLL algorithm

Determine if an input propositional logic sentence (in CNF) is satisfiable. **This is just backtracking search for a CSP.**

Improvements:

1. Early termination

A clause is true if any literal is true.

A sentence is false if any clause is false.

2. Pure symbol heuristic

Pure symbol: always appears with the same "sign" in all clauses.

e.g., In the three clauses  $(A \vee \neg B)$ ,  $(\neg B \vee \neg C)$ ,  $(C \vee A)$ , A and B are pure, C is impure.

Make a pure symbol literal true. (if there is a model for S, then making a pure symbol true is also a model).

3 Unit clause heuristic

Unit clause: only one literal in the clause

The only literal in a unit clause must be true.

Note: literals can become a pure symbol or a unit clause when other literals obtain truth values. e.g.

$$(\cancel{A \vee \text{True}}) \wedge (\neg A \vee B)$$

$A = \text{pure}$

# The WalkSAT algorithm

- Incomplete, local search algorithm
- Evaluation function: The min-conflict heuristic of minimizing the number of unsatisfied clauses
- Balance between greediness and randomness

## Walksat Procedure

Start with random initial assignment.

Pick a random unsatisfied clause.

Select and flip a variable from that clause:

With probability  $p$ , pick a **random** variable.

With probability  $1-p$ , pick **greedily**

a variable that minimizes the number of unsatisfied clauses

Repeat to predefined maximum number flips;  
if no solution found, restart.

# Hard satisfiability problems

- Consider *random* 3-CNF sentences. e.g.,  
 $(\neg D \vee \neg B \vee C) \wedge (B \vee \neg A \vee \neg C) \wedge (\neg C \vee \neg B \vee E) \wedge (E \vee \neg D \vee B) \wedge (B \vee E \vee \neg C)$

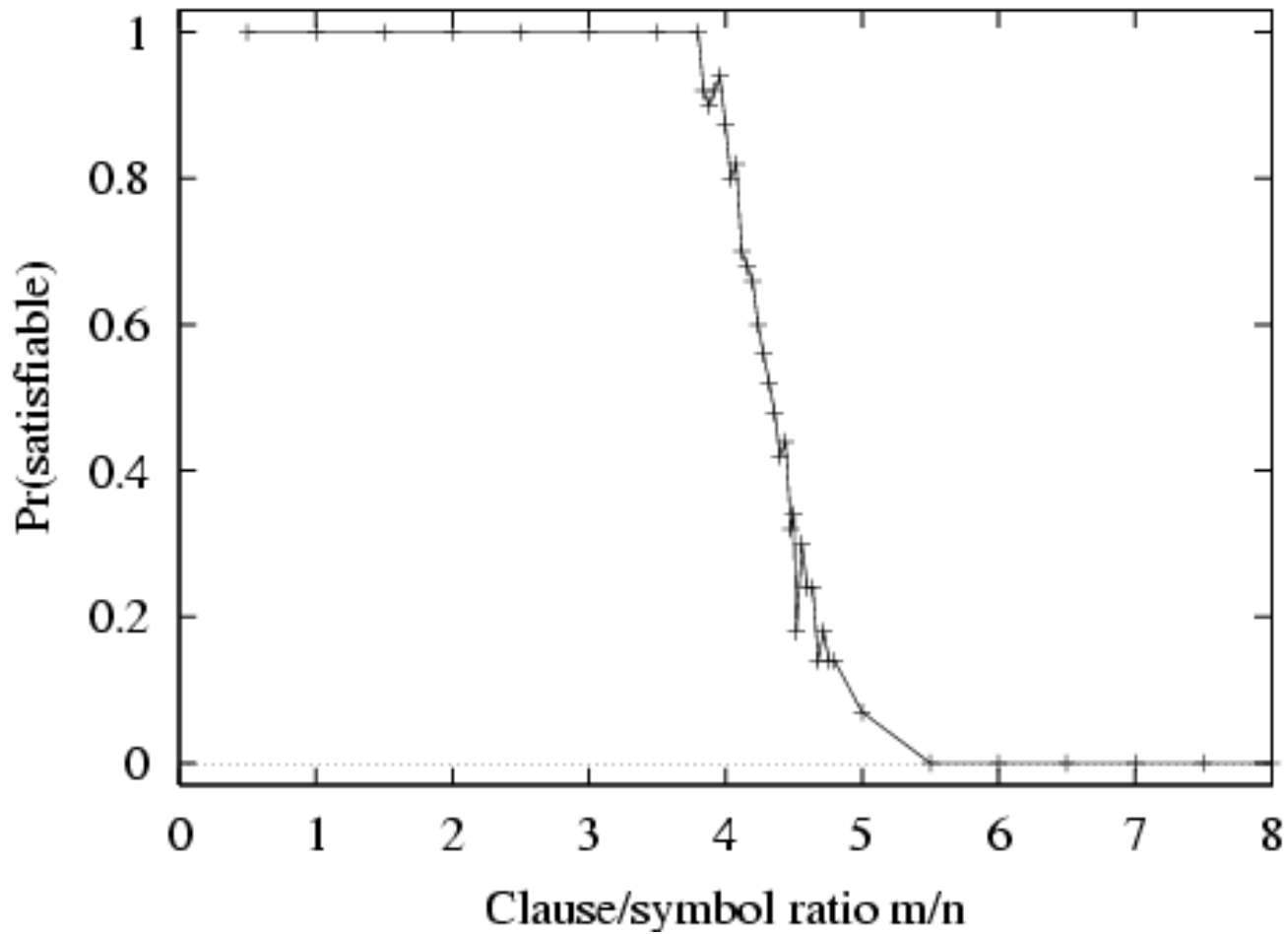
$m$  = number of clauses (5)

$n$  = number of symbols (5)

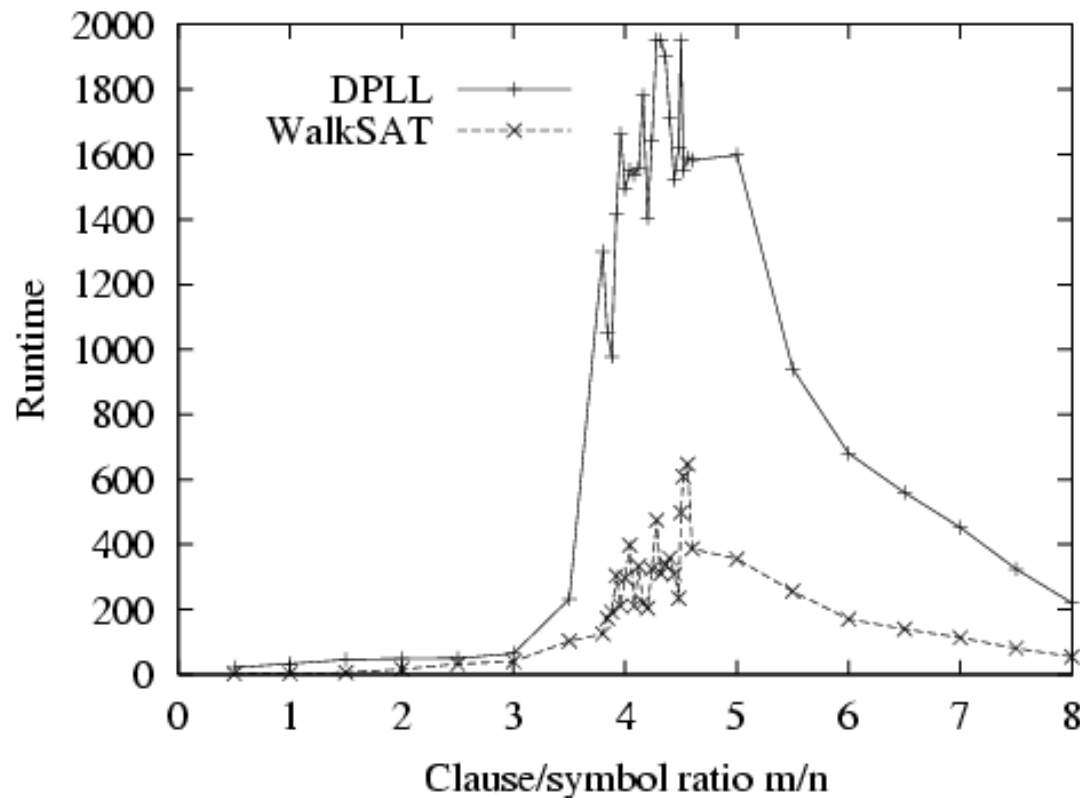
- Hard problems seem to cluster near  $m/n = 4.3$   
(critical point)



# Hard satisfiability problems



# Hard satisfiability problems



- Median runtime for 100 **satisfiable** random 3-CNF sentences,  $n = 50$

# Common Sense Reasoning

Example, adapted from Lenat

You are told: John drove to the grocery store and bought a pound of noodles, a pound of ground beef, and two pounds of tomatoes.

- Is John 3 years old?
  - Is John a child?
  - What will John do with the purchases?
  - Did John have any money?
  - Does John have less money after going to the store?
  - Did John buy at least two tomatoes?
  - Were the tomatoes made in the supermarket?
  - Did John buy any meat?
  - Is John a vegetarian?
  - Will the tomatoes fit in John's car?
- 
- Can Propositional Logic support these inferences?

# 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
- Resolution is complete for propositional logic.  
Forward and backward chaining are linear-time, complete for Horn clauses
- Propositional logic lacks expressive power