

(Chapter-7) Logical Agents---Resolution

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From KB. derive α from [KB,- -a), derive False [KB, a) is unsatisfiable using resolution inference rule

☐ is the following inference rule sound?

$$\frac{AV \neg B, BVC}{AVC}$$

■ Resolution rule

Complementary Literals [7.5.2)

CNF (7.5.2)

□ Clause
 disjunction of literals
 E.g., AV¬B
 □ Conjunctive Normal Form
 conjunction of disjunctions of literals
 E.g.,(AV¬B)Λ(BVC):

Basic intuition, resolve B, ¬B to AVC

CNF:why?how?

Converting wffs to Conjunctions of Clauses

- •Resolution is a powerful tool for algorithmic inference, but we can only apply it to conjunctions of clauses (conjunctive normal form, CNF).
- @Fortunately, there is such a way, allowing us to apply resolution to any wff.

Conversion to CNF

Conjunctive Normal Form (CNF)

conjunction of disjunctions of literals: clauses

E.g.,
$$(A \lor \neg B) \land (B \lor \neg C \lor \neg D)$$

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

1. Eliminate \Leftrightarrow , replacing $\alpha \Leftrightarrow \beta$ with $(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$.

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

2. Eliminate \Rightarrow , replacing $\alpha \Rightarrow \beta$ with $\neg \alpha \lor \beta$.

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

3. Move ¬ inwards using de Morgan's rules and double-negation:

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

4. Apply distributivity law (∧ over ∨) and flatten:

$$(\neg B_{1.1} \lor P_{1.2} \lor P_{2.1}) \land (\neg P_{1.2} \lor B_{1.1}) \land (\neg P_{2.1} \lor B_{1.1})$$

Converting wffs to Conjunctions of Clauses

- **@Example:** $\neg (P \rightarrow Q) \lor (R \rightarrow P)$.
- Step 1: Eliminate implication operators:
- $@\neg(\neg P\lor Q)\lor(\neg R\lor P)$
- **Step 2:** Reduce the scopes of \neg operators by using DeMorgan's laws and eliminating double \neg operators:
- $@(P \land \neg Q) \lor (\neg R \lor P)$
- $@(P \lor \neg R \lor P) \land (\neg Q \lor \neg R \lor P)$, and then
- $@(P \lor \neg R) \land (\neg Q \lor \neg R \lor P)$

- •Resolution is a sound rule of inference, but it is not complete.
- **©** For example, $P \wedge R \neq P \vee R$, but the resolution rule does not allow us to infer $P \vee R$ from $\{P\}$, $\{R\}$.
- **@**However, we can use resolution to show that the negation of $P \lor R$ is inconsistent with $\{P\}$, $\{R\}$ and thereby showing that $P \land R \models P \lor R$.
- **@**The negation of $P \vee R$ is $\neg P \wedge \neg R$.
- **©**Conjunctively combining all clauses results in $\neg P \land \neg R \land P \land R$.
- **Q**This resolves to the empty set, so by contradiction we have indirectly shown that $P \wedge R \neq P \vee R$.

More Explanation of Resolution Refutations

- **Q**You have a set of hypotheses h_1 , h_2 , ..., h_n , and a conclusion c.
- **Q**Your argument is that whenever all of the h_1 , h_2 , ..., h_n are true, then c is true as well.
- QIn other words, whenever all of the h_1 , h_2 , ..., h_n are true, then $\neg c$ is false.
- **Q**If and only if the argument is valid, then the conjunction $h_1 \wedge h_2 \wedge ... \wedge h_n \wedge \neg c$ is false, because either (at least) one of the h_1 , h_2 , ..., h_n is false, or if they are all true, then $\neg c$ is false.
- Therefore, if this conjunction resolves to false, we have shown that the argument is valid.

Resolution

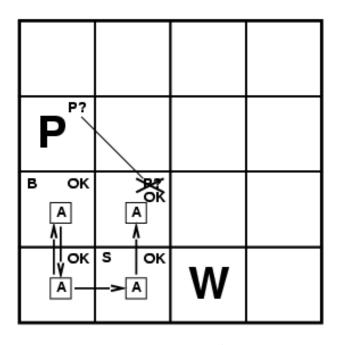
• Resolution inference rule (for CNF):

$$\frac{I_1 \vee ... \vee I_k}{I_1 \vee ... \vee I_{i-1} \vee I_{i+1} \vee ... \vee I_k \vee m_1 \vee ... \vee m_{j-1} \vee m_{j+1} \vee ... \vee m_n}$$

where I_i and m_i are complementary literals.

E.g.,
$$P_{1,3} \vee P_{2,2}$$
, $\neg P_{2,2}$

Resolution is sound and complete for propositional logic



Resolution

$$\frac{P, \neg P \lor Q}{Q}$$

$$\frac{P \vee Q, \neg P \vee Q}{Q}$$

$$\frac{P, \neg P}{NIL}$$

☐ Sound?

\square Sound?

- \square How to prove KB $= \alpha$ using resolution?
- \square consider KB $\models \alpha$ iff (KB $\land \neg \alpha$) is unsatisfiable and

$$\frac{P - P}{NIL}$$

@Example:

- @"Gary is intelligent or a good actor.
- If Gary is intelligent, then he can count from 1 to 10.
- @Gary can only count from 1 to 2.
- ©Therefore, Gary is a good actor."

@Propositions:

- @I: "Gary is intelligent."
- QA: "Gary is a good actor."
- ©C: "Gary can count from 1 to 10."

@Hypotheses:

$$I \vee A, I \rightarrow C, \neg C$$

In CNF: $(I \lor A) \land (\neg I \lor C) \land \neg C$

- @Conclusion: A
- @Conjunction of Clauses for Resolution Refutation:
- $@(I \lor A) \land (\neg I \lor C) \land \neg C \land \neg A$
- **@Resolution on A:** $I \land (\neg I \lor C) \land \neg C$
- **@Resolution on C: False**
- **@**Therefore, the initial set of clauses is inconsistent, and the conclusion is correct: Gary is a good actor.

@Another Example:

- @"If Jim visits a pub on Thursday, he is late for work on Friday.
- @If Jim is late for work on Friday, he has to work during the weekend.
- @Jim had to work during the weekend.
- ©Therefore, Jim visited a pub on Thursday."

@Propositions:

@T: "Jim visits a pub on Thursday."

@F: "Jim is late for work on Friday."

@W: "Jim has to work during the weekend."

- **@Hypotheses:**
- $QT \supset F, F \supset W, W$
- $@In CNF: (\neg T \lor F) \land (\neg F \lor W) \land W$
- @Conclusion:T
- @Conjunction of Clauses for Resolution Refutation:
- $@(\neg T \lor F) \land (\neg F \lor W) \land W \land \neg T$
- @Resolution on F:
- $@(\neg T \lor W) \land W \land \neg T$
- **@Simplification:**
- **@W ∧ ¬T**
- One of the control of the co
- Therefore, the argument is invalid and the conclusion that Jim went to a
 pub on Thursday is incorrect.
- **@**(He could have been forced to work during the weekend for other reasons.)

Resolution algorithm

@Proof by contradiction, i.e., show $KB \land \neg \alpha$ **unsatisfiable**

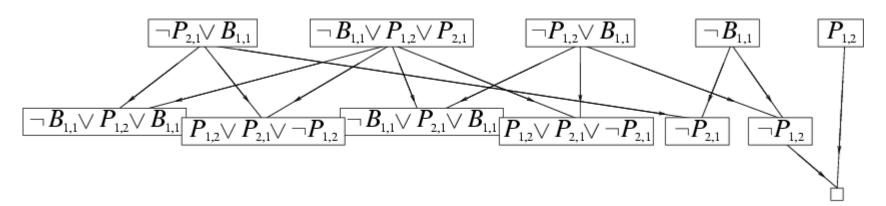
```
function PL-RESOLUTION(KB, a)returns true or false
    clauses ← the set of clauses in the CNF representation of KB ∧ ¬a
    new ←{ }
    loop do
        for each C<sub>i</sub>, C<sub>j</sub> in clauses do
            resolvents ← PL-RESOLVE(C<sub>i</sub>, C<sub>j</sub>)
        if resolvents contains the empty clause then return true
            new←new ∪ resolvents
        if new□ clauses then return false
        clauses ← clauses ∪ new
```

Resolution example

$$@KB = (B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1})) \land \neg B_{1,1}$$

$$Q \alpha = \neg P_{1,2}$$

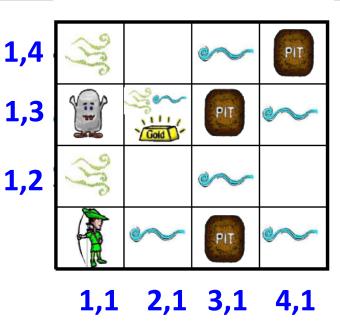
@
$$(\neg P_{2,1} \lor B_{1,1}) \land (\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg P_{1,2} \lor B_{1,1})$$



If a set of clauses is unsatisfiable, then the resolution closure of those clausescontains the empty clause.

Logical agent in the Wumpus world

- ☐ Start
 - $R_1:\neg P_{1.1}$
- ☐ Pits cause breezes in adjacent squares
 - $\mathbb{R}_2: \mathbb{B}_{1,1} <=> (\mathbb{P}_{1,2} \mathbb{V} \mathbb{P}_{2,1})$
 - \mathbb{I} R₃:B_{2,1}<=>(P_{1,1}VP_{2,2}VP_{3,1})
- Perceiving
 - $R_4:\neg B_{1,1}$
 - $R_5:B_{2,1}$
- \square KB $\models \neg P_{1,2}$?
- ☐ CNF representation of KB ...
- ☐ resolution ...



{PVQ,¬PVQ,PV¬Q,¬PV¬Q}

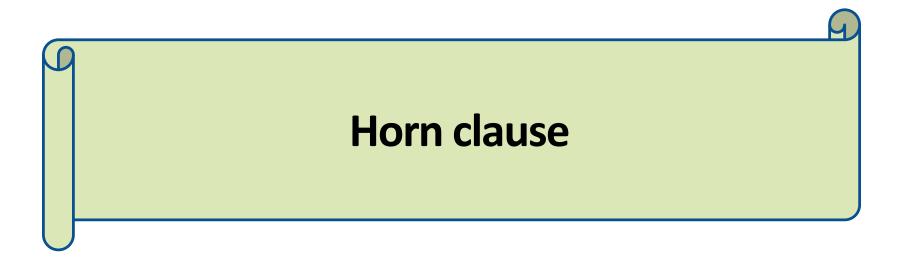
is unsatisfiable

```
1) PVQ
2) ¬PVQ
3) PV¬Q
4) ¬PV¬Q
5) Q 1,2
6) P 1,3
7) PV¬P 1,4
8) QV¬Q 1,4
9) QV¬Q 2,3
10) PV¬P 2,3
11) ¬P 2,4
12) ¬Q 3,4
```

```
14) PVQ
       1,8
15) PVQ 1,9
16) PVQ 1,10
17) Q
       1,11
18) P 1,12
19) Q 2,6
20) ¬PVQ 2,7
21) ¬PVQ 2,8
22) ¬PVQ 2,9
23) ¬PVQ 2,10
24) ¬P 2,12
25) P 3,5
26) PV¬Q 3,7
```

```
27) PV¬Q 3,8
28) PV¬Q 3,9
29) PV¬Q 3,10
30) ¬Q 3,11
31) ¬P 4,5
32) ¬Q 4,6
33) ¬PV¬Q 4,7
34) ¬PV¬Q 4,8
35) ¬PV¬Q 4,9
36) ¬PV¬Q4,10
         5,8
37) Q
38) Q 5,9
39) NIL 5,12
```

think: how to improve the efficiency?



Horn clause, Definite clause (7.5.3)

- One restricted form is Definite clause, which is a
 disjunction of literals of which exactly one is positive
 - 1. $(\neg P_{1,1} \lor \neg B_{1,1} \lor S_{1,1})$ is a definite clause
 - 2. $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1})$ is not.
- Slightly more general is the Horn clause, which is a disjunction of literals of which at most one is positive.
- QAll definite clauses are Horn clauses, as are clauses with on positive literals, these are called goal clauses.
- @Horn clauses are closed under resolution: if you resolve two Horn clauses, you get back a Horn clause.

Horn clause, Definite clause [7.5.3]

Every definite clause can be written as an implication whose premise is a conjunction of positive literals and whose conclusion is a single positive literal.

```
\neg L_{1,1} \lor \neg Breeze \lor B_{1,1} can be expressed as L_{1,1} \land Breeze \Rightarrow B_{1,1}
```

- Inference with Horn clauses can be done through the forward-chaining and backward chaining algorithms.
- © Deciding entailment with Horn clauses can be done in time that is linear in the size of the knowledge base



Forward and backward chaining

- Phorn Form (restricted)
 - 1. KB = conjunction of Horn clauses
- @Horn clause : a disjunction of literals of which at most one is positive
 - E.g., $\neg L_{1,1} \lor \neg$ Breeze $\lor B_{1,1}$, $L_{1,1} \land$ Breeze $\Rightarrow B_{1,1}$
 - proposition symbol, (conjunction of symbols) ⇒ symbol
 - E.g., $C \land (B \Rightarrow A) \land (C \land D \Rightarrow B)$,
- Modus Ponens (for Horn Form): complete for Horn KBs

$$\alpha_1, ..., \alpha_n,$$
 $\alpha_1 \wedge ... \wedge \alpha_n \Rightarrow \beta$

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

Forward chaining {7.5.4}

```
while agenda is not empty do
    p←POP( agenda)
    if p = q then return true
    if inferred[p] = false then
        inferred[p] ← true
        for each clause c in KB where p is in c.PREMISE do
            decrement count[ c]
        if count[c] = 0 then add c.CONCLUSION to agenda
return false
```

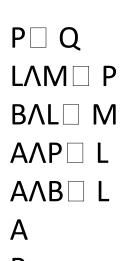
P□ Q LΛM□ P BΛL□ M AΛP□ L AΛB□ L A

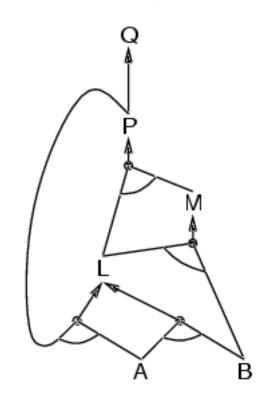
Forward chaining {7.5.4}

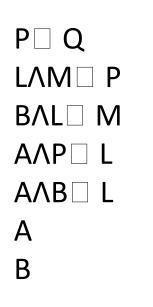
- **The** Agenda keeps track of symbols known to be true but not yet 'processed'.
- The count table keeps track of how many premises of each implication are as yet unknown. Whenever a new symbol p from the agenda is processed, the count is reduced by one for each implication in whose premise p appears. If a count reaches zero, all the premises of the implication are known, so it is conclusion can be added to the agenda.
- Forward chaining is an example of data-driven reasoning:
 - reasoning in which the focus of attention starts with the known data.

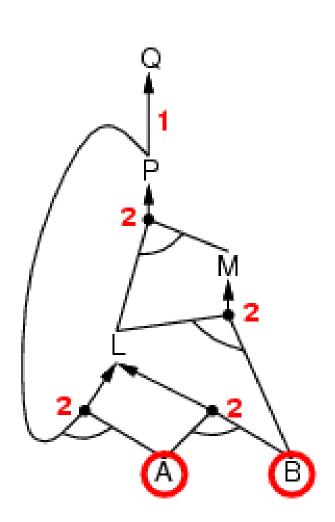
Forward chaining

- @Idea: find any rule whose premises are satisfied in the KB,
 - 1. add its conclusion to the KB, until query is found

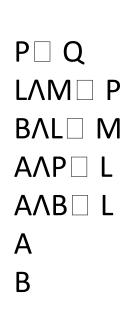


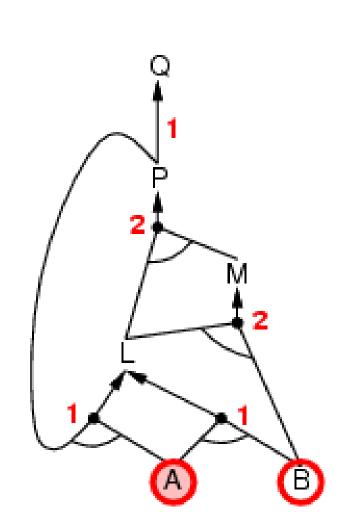




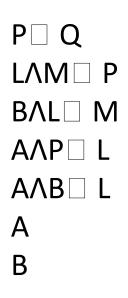


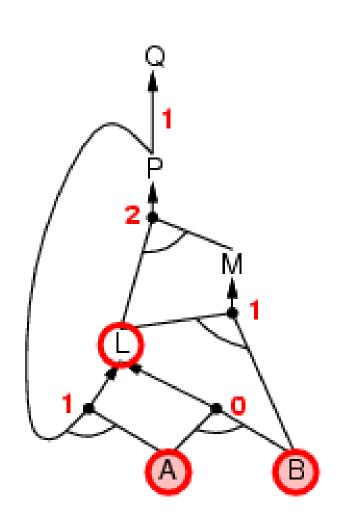
A B



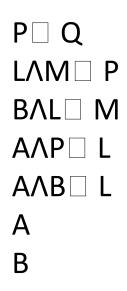


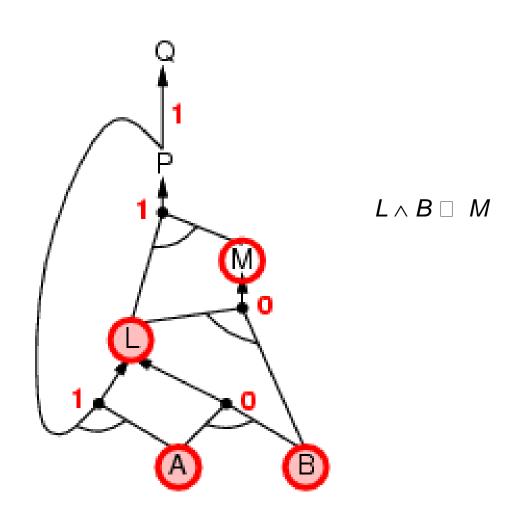
 $A \wedge ?P \square ?L$

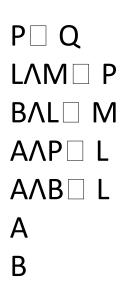


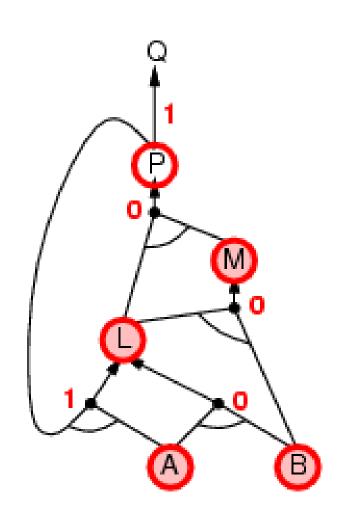


 $A \wedge B \square L$

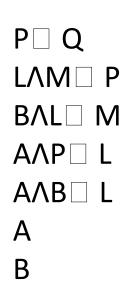


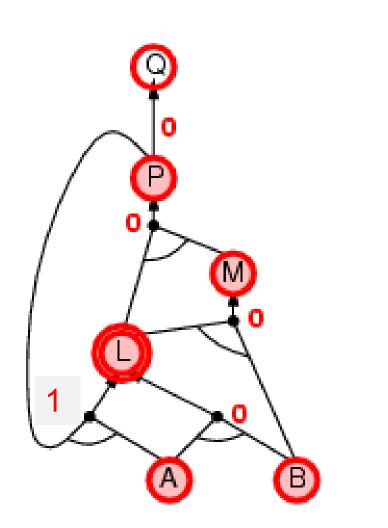




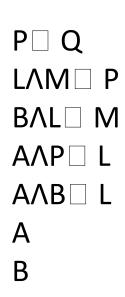


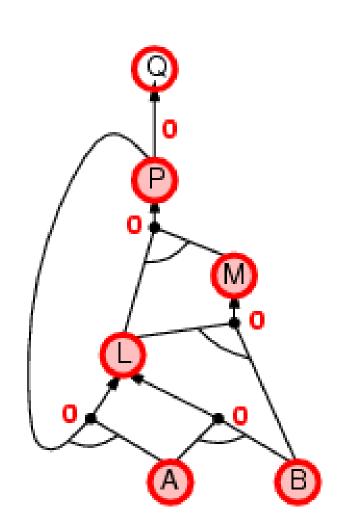
 $M \wedge L \square P$



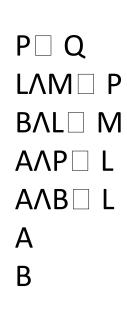


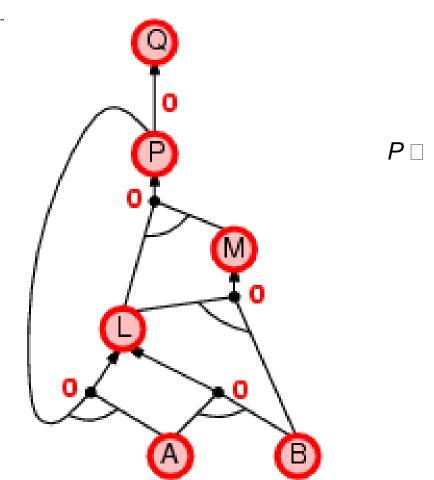
 $P \square Q$





 $A \wedge P \square L$





 $P \square Q$

Proof of completeness

FC derives every atomic sentence that is entailed by KB

- FC reaches a fixed point where no new atomic sentences are derived
- Consider the final state as a model m, assigning true/false to symbols
- 3. Every clause in the original *KB* is true in *m*

$$a_1 \wedge ... \wedge a_k \Rightarrow b$$

- 4. Hence *m* is a model of *KB*
- 5. If $KB \models q, q$ is true in every model of KB, including m



Backward chaining{7.5.4}

- •Works backward from the query. If the query q is known to be true, then no work is needed.
 Otherwise, the algorithm finds those implications in the knowledge base whose conclusion is q. If all the premises of one of those implications can be proved true(by backward chaining), then q is true.
- @Backward chaining is a form of goal-directed reasoning.

Backward chaining

Idea: work backwards from the query q:

```
to prove q by BC,

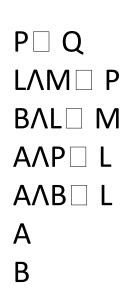
check if q is known already, or

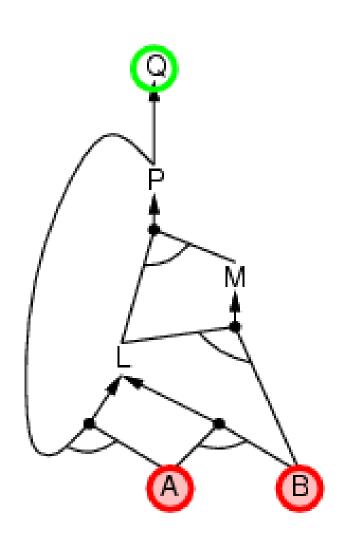
prove by BC all premises of some rule concluding q
```

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

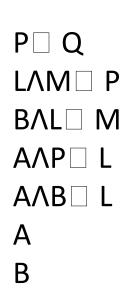
Avoid repeated work: check if new sub-goal

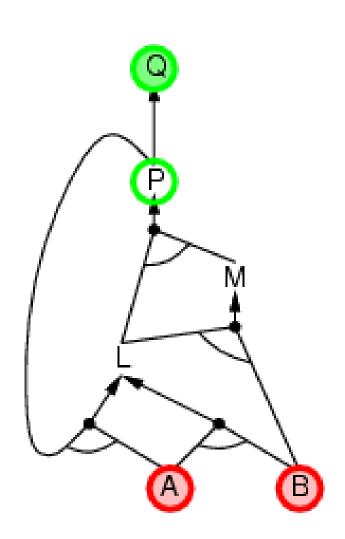
- 1. has already been proved true, or
- 2. has already failed



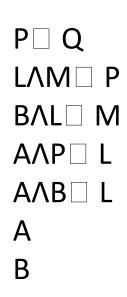


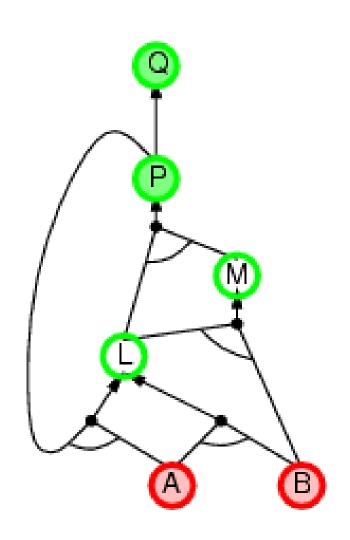
A B Q?

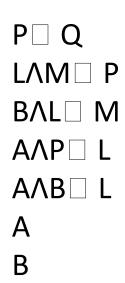


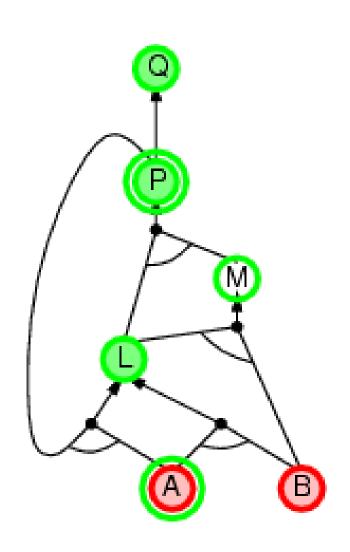


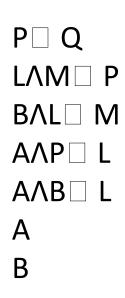
A B Q?<=P?

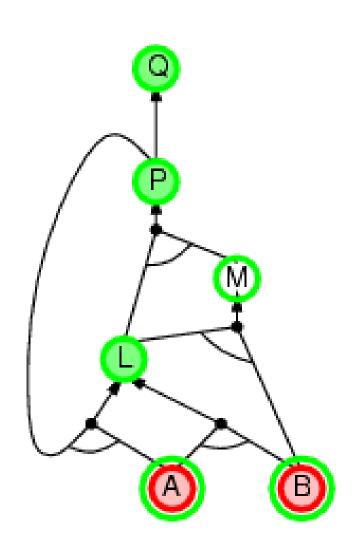




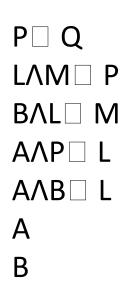


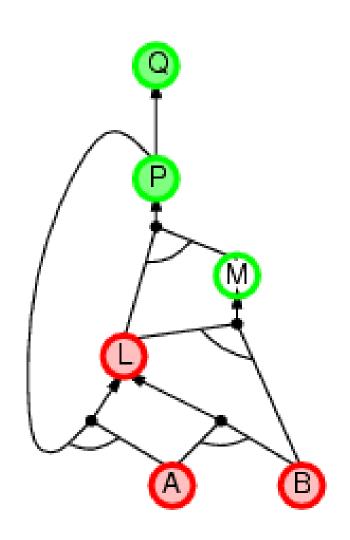




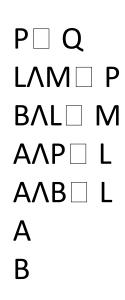


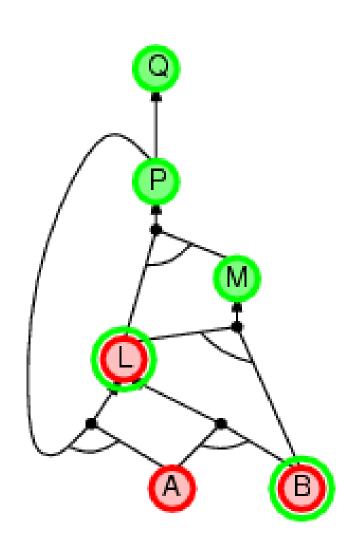
A
B
$$Q? <= P?$$
 $L? \land M? => P?$
 $P? \land A => L?$
 $A \land B => L$



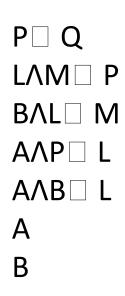


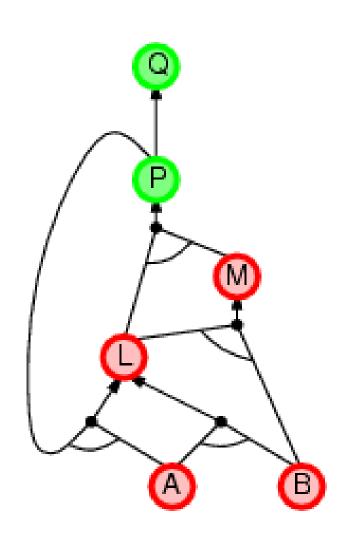
A
B
$$Q? <= P?$$
 $L? \land M? => P?$
 $P? \land A => L?$
 $A \land B => L$



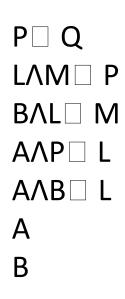


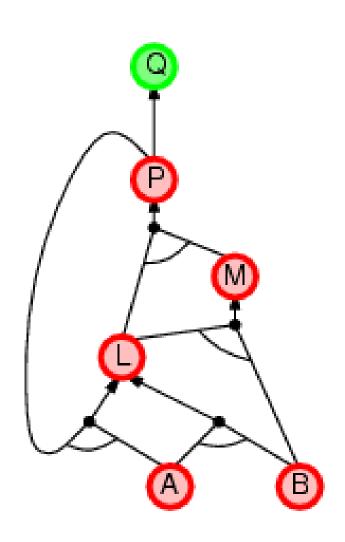
A
B
$$Q? <= P?$$
 $L? \land M? => P?$
 $P? \land A => L?$
 $A \land B => L$
 $L \land B => M$



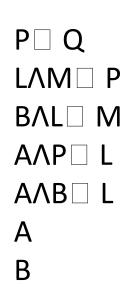


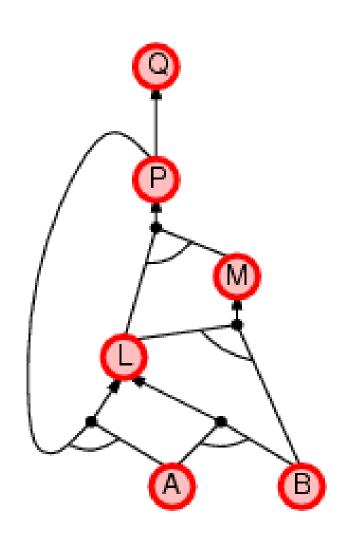
A
B
$$Q? <= P?$$
 $L \land M => P?$
 $P? \land A => L?$
 $A \land B => L$
 $L \land B => M$





A
B
$$Q? <= P$$
 $L \land M => P$
 $P? \land A => L?$
 $A \land B => L$
 $L \land B => M$





A B Q <= P $L \land M => P$ $P \land A => L$ $A \land B => L$ $L \land B => M$

Forward vs. backward chaining

- @FC is data-driven, automatic, unconscious processing,
 - 1. 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,
 - 1. 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

Sumary

- □ 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 with regard to

models

- Propositional logic lacks expressive power
 - can't say " pits cause breezes in adjacent squares "
 - except by writing one sentence for each square

Expressiveness limitation of propositional logic

- **@** For every time t and every location $[x^t, y^t]$,

$$L_{x,y}^{t} \wedge FacingRight^{t} \wedge Forward^{t} \Rightarrow L_{x+1,y}^{t+1}$$

Rapid proliferation of clauses

Summary

- Logical agents apply inference to a knowledge base to derive new information and make decisions
- @Basic concepts of logic:
 - 1. syntax: formal structure of sentences
 - 2. semantics: truth of sentences wrt models
 - 3. entailment: necessary truth of one sentence given another
 - 4. inference: deriving sentences from other sentences
 - 5. soundness: derivations produce only entailed sentences
 - 6. completeness: derivations can produce all entailed sentences

Summary

- Wumpus world requires the ability to represent partial and negated information, reason by cases, etc.
- Resolution is complete for propositional logic
- © Forward, backward chaining are linear-time, complete for Horn clauses
- Propositional logic lacks expressive power

Propositional Logic (PL)

- @P.L. offers techniques for capturing facts or rules in symbolic form and then operates on them through use of logical operators.
- PL provide method of managing statements that are either TRUE or FALSE.
- Prolog is based on PC

Propositional Calculus

- If we build a machine that represents its knowledge as propositions, we can use these mechanisms to enable the machine to deduce new knowledge from existing knowledge and verify hypotheses about the world.
- @However, propositional calculus has some serious restrictions in its capability to represent knowledge.

@

Propositional Calculus

- In propositional calculus, atoms have no internal structure; we cannot reuse the same proposition for a different object, but each proposition always refers to the same object.
- @For example, in the toy block world, the propositions ON_A_B and ON_A_C are completely different from each other.
- **@**We could as well call them PETER and BOB instead.
- **Q**So if we want to express rules that apply to a whole class of objects, in propositional calculus we would have to define separate rules for every single object of that class.

Predicate Calculus

- **@This leads us to predicate calculus in next lecture.**
- Predicate calculus has symbols called
- object constants,
- relation constants, and
- function constants
- These symbols will be used to refer to objects in the world and to propositions about the word.





Thank you

End of
Chapter 7-2