

#### **Outline**

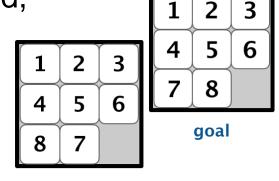
- Knowledge-based agents
- The Wumpus world
- Inference with Propositional logic
- Propositional theorem proving
- Effective propositional model checking (Self-study)

## Knowledge-based agents

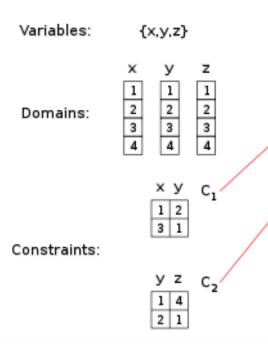
#### Problem-solving agents

- These agents know things in a very limited, inflexible sense.
  - E.g., an 8-puzzle agent cannot deduce pairs of unsolvable states from their parities.

{C<sub>1</sub> C<sub>2</sub>}



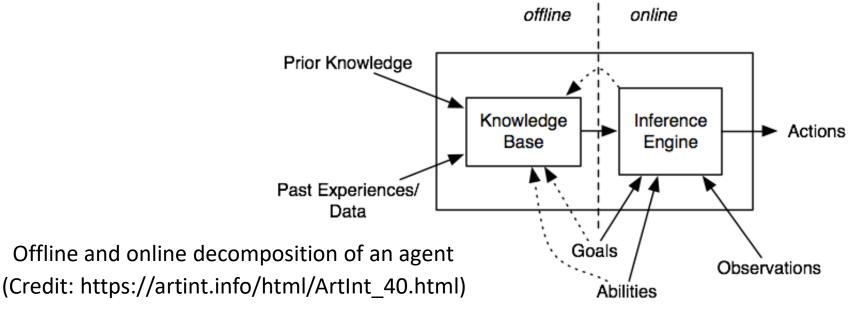
unsolvable



- CSP enables some parts of the agent to work domain-independently
  - State = an assignment of values to variables
  - Allow for more efficient algorithms

#### Knowledge-based agents

- Supported by logic a general class of representation
- Combine and recombine information to suit myriad purposes
  - Accept new tasks in the form of explicitly described goals
  - Achieve competence by learning new knowledge of the environment
  - Adapt to changes by updating the relevant knowledge



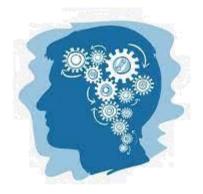
#### Knowledge-based agents



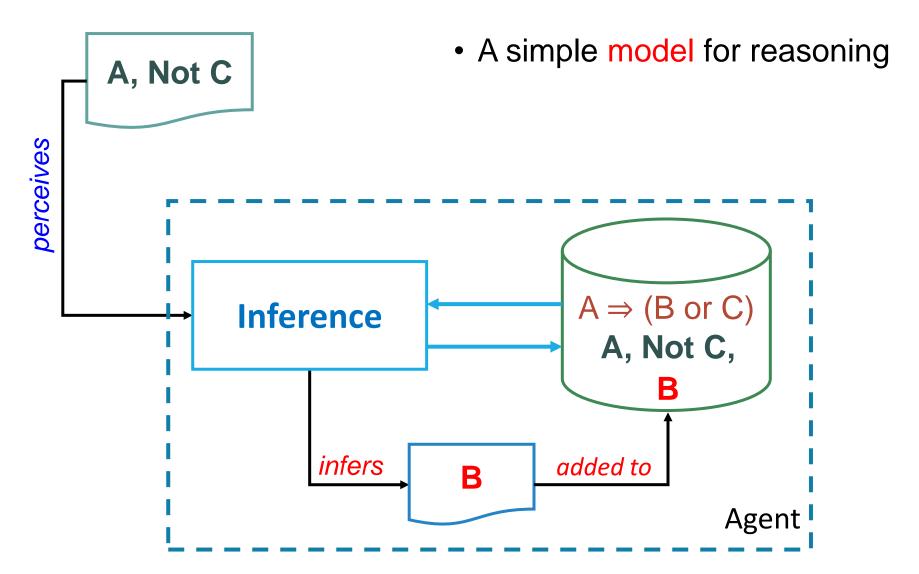


- Each sentence represents some assertion about the world.
- Axiom = sentence that is not derived from other sentences

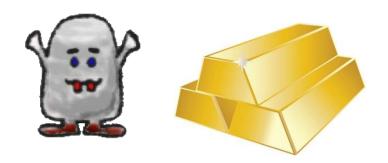
- Inference: Derive (infer) new sentences from old ones
  - Add new sentences to the knowledge base and query what is known



#### Model for reasoning: An example



# The Wumpus world



SSSSS SSSSSSSSSSSSSSSSSSSSSSSSSSSSSSSS		Breeze	PIT
200	Breeze SSSSSS Stench Gold	PIT	Breeze
SSSSSS SStench		Breeze	
START	Breeze	PIT	Breeze

#### PEAS Description

#### Environment

- 4×4 grid of rooms, agent starts in the square [1,1], facing to the right
- The locations of Gold and Wumpus are random
- Each square can be a pit, with probability 0.2

#### Performance measure

- +1000 for climbing out of the cave with gold, -1000 for death
- -1 per step, -10 for using the arrow
- The game ends when agent dies or climbs out of the cave
- Actuators: Forward, TurnLeft/TurnRight by 90°, Grab, Shoot, Climb
- Sensors: Stench, Breeze, Glitter, Bump, Scream
- Percept: [Stench, Breeze, None, None, None]

A = Agent

 $\mathbf{B} = Breeze$ 

**G** = Glitter, Gold

OK = Safe square

P = Pit

S = Stench

V = Visited

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2 OK	2,2	3,2	4,2
1,1 A	2,1	3,1	4,1
OK	OK		

A = Agent

 $\mathbf{B} = Breeze$ 

G = Glitter, Gold

OK = Safe square

P = Pit

S = Stench

V = Visited

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2 OK	2,2 <b>P?</b>	3,2	4,2
1,1 V OK	2,1 A B OK	3,1 P?	4,1

A = Agent

 $\mathbf{B} = Breeze$ 

G = Glitter, Gold

OK = Safe square

P = Pit

S = Stench

V = Visited

1,4	2,4	3,4	4,4
1,3 W!	2,3	3,3	4,3
1,2 S OK	2,2 OK	3,2	4,2
1,1 V OK	2,1 B V OK	3,1 P!	4,1

A = Agent

 $\mathbf{B} = Breeze$ 

G = Glitter, Gold

OK = Safe square

P = Pit

S = Stench

V = Visited

1,4	2,4 <b>P</b> ?	3,4	4,4
1,3 <b>W!</b>	2,3 A S G B	3,3 <sub>P?</sub>	4,3
1,2 s	2,2	3,2	4,2
${f V}$	$\mathbf{v}$		
OK	OK		
1,1	2,1 B	3,1 <b>P!</b>	4,1
V	V		
OK	OK		

#### An agent in the Wumpus world

 A Wumpus-world agent using propositional logic will have a KB of 64 distinct proposition symbols, 155 sentences.

$$\begin{array}{l} \neg P_{1,1} \\ \neg W_{1,1} \\ B_{x,y} \Leftrightarrow (P_{x,y+1} \vee P_{x,y-1} \vee P_{x+1,y} \vee P_{x-1,y}) \\ S_{x,y} \Leftrightarrow (W_{x,y+1} \vee W_{x,y-1} \vee W_{x+1,y} \vee W_{x-1,y}) \\ W_{1,1} \vee W_{1,2} \vee \ldots \vee W_{4,4} \\ \neg W_{1,1} \vee \neg W_{1,2} \\ \neg W_{1,1} \vee \neg W_{1,3} \\ \ldots \end{array}$$

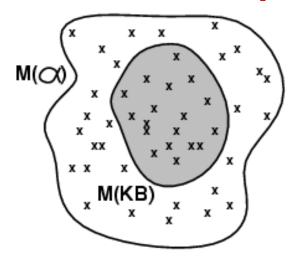
# Propositional logic

#### Logics in general

- Models (or possible worlds) are mathematical abstractions that fix the truth or falsehood of every relevant sentence.
  - E.g., all possible assignments of real numbers to x and y
- m satisfies (or is a model of)  $\alpha$  if  $\alpha$  is true in model m
- $M(\alpha)$  = the set of all models of  $\alpha$

#### Entailment in logic

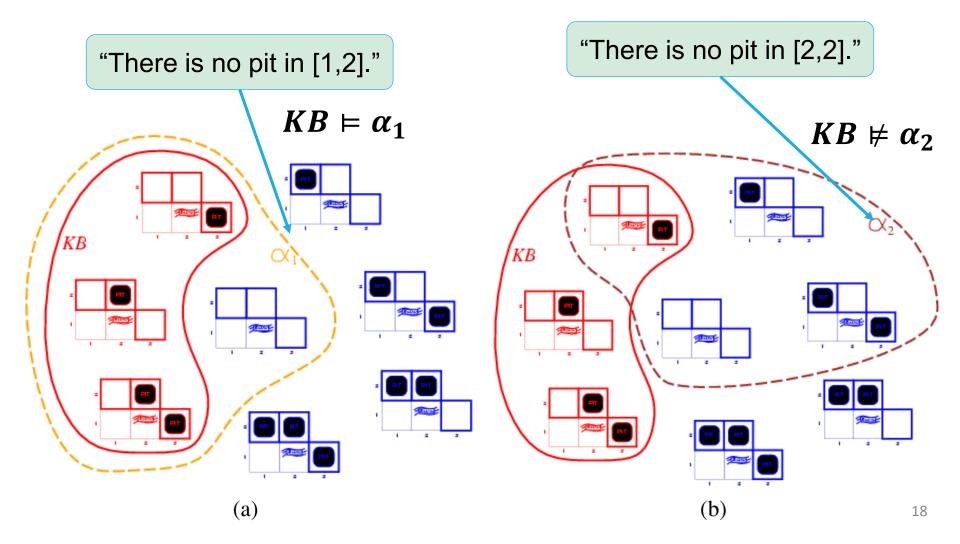
- A sentence follows logically from another sentence:  $\alpha \models \beta$
- $\alpha \vDash \beta$  if and only if, in every model in which  $\alpha$  is true,  $\beta$  is also true, i.e.,  $M(\alpha) \subseteq M(\beta)$



- For example,
  - x = 0 entails xy = 0
  - The KB containing "Apple is red" and "Tomato is red" entails "Either the apple or the tomato is red"
- Entailment is a relationship between sentences (i.e., syntax) that is based on semantics.

#### Entailment in logic: Wumpus world

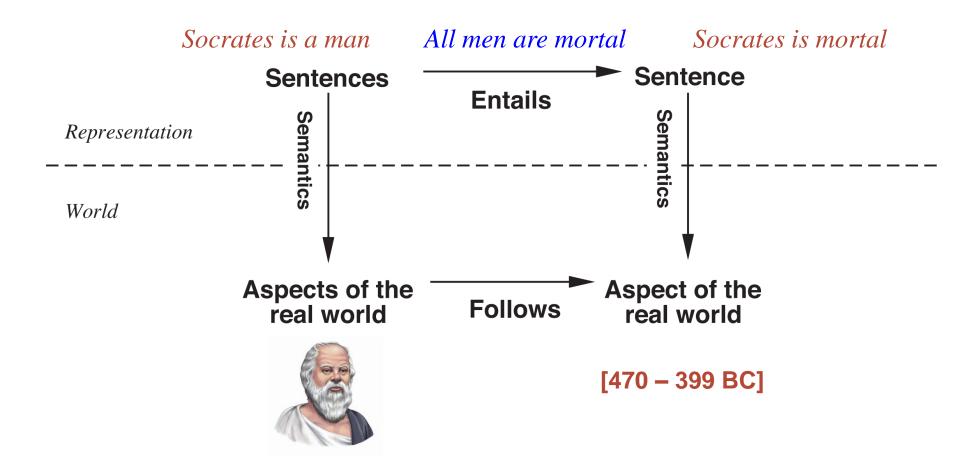
• Consider two possible conclusions  $\alpha_1$  and  $\alpha_2$ 



#### Logical inference

- $KB \models_i \alpha$  means  $\alpha$  can be derived from KB by procedure i
- Soundness: i is sound if whenever  $KB \models_i \alpha$ , it is also true that  $KB \models \alpha$
- Completeness: i is complete if whenever  $KB \models \alpha$ , it is also true that  $KB \models_i \alpha$
- That is, the procedure will answer any question whose answer follows from what is known by the KB.

#### World and representation



### Propositional logic: Syntax

#### Propositional logic: Syntax

```
Sentence \rightarrow AtomicSentence \mid ComplexSentence
 AtomicSentence \rightarrow True \mid False \mid P \mid Q \mid R \mid \dots
ComplexSentence \rightarrow (Sentence) \mid [Sentence]
                            \neg Sentence
                          Sentence \land Sentence
                           Sentence \lor Sentence
                          Sentence \Rightarrow Sentence
                           Sentence \Leftrightarrow Sentence
```

OPERATOR PRECEDENCE :  $\neg, \land, \lor, \Rightarrow, \Leftrightarrow$ 

Literal: atomic sentence (P) or negated atomic sentence ( $\neg$ P)

#### Propositional logic: Semantics

- Each model specifies true/false for each proposition symbol.
  - E.g.,  $m_1 = \{P_{1,2} = false, P_{2,2} = false, P_{3,1} = true\}$ , 8 possible models
- Rules for evaluating truth with respect to a model m

P	Q	$\neg P$	$P \wedge Q$	$P \lor Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
false false true true	false true false true	$true \ true \ false \ false$	$false \\ false \\ false \\ true$	$false \ true \ true \ true$	$true \ true \ false \ true$	$true \\ false \\ false \\ true$

- Simple recursive process evaluates an arbitrary sentence.
  - E.g.,  $\neg P_{1,2} \land (P_{2,2} \lor P_{3,1}) = true \land (true \lor false) = true \land true = true$

#### A simple knowledge base

- Symbols for each position [i, j]
  - $P_{i,j}$ : there is a pit in [i,j]
  - $W_{i,j}$ : there is a Wumpus in [i,j]
- $B_{i,j}$ : there is a breeze in [i,j]
- $S_{i,j}$ : there is a stench in [i,j]
- Sentences in Wumpus world's KB

$$R_1$$
:  $\neg P_{1,1}$ 
 $R_2$ :  $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$ 
 $R_3$ :  $B_{2,1} \Leftrightarrow (P_{1,1} \vee P_{2,2} \vee P_{3,1})$ 
 $R_4$ :  $\neg B_{1,1}$ 
 $R_5$ :  $B_{2,1}$ 

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2 OK	2,2 P?	3,2	4,2
1,1 V OK	2,1 A B OK	3,1 P?	4,1

#### A simple inference procedure

- Given: a set of sentences, KB, and sentence  $\alpha$
- Goal: answer  $KB \models \alpha$ ? = "Does KB semantically entail  $\alpha$ ?"
  - In all interpretations in which KB's sentences are true, is  $\alpha$  also true?
  - E.g., in the Wumpus world,  $KB = P_{1,2}$ ? = "Is there a pit in [1,2]?"

Model-checking approach (Inference by enumeration)

Inference rules

Conversion to the inverse SAT problem (Resolution refutation)

#### Model-checking approach

- Check if  $\alpha$  is true in every model in which KB is true.
  - E.g., the Wumpus's KB has 7 symbols  $\rightarrow 2^7 = 128$  models
- Draw a truth table for checking

No pit in [1,2]

$B_{1,1}$	$B_{2,1}$	$P_{1,1}$	$P_{1,2}$	$P_{2,1}$	$P_{2,2}$	$P_{3,1}$	$R_1$	$R_2$	$R_3$	$R_4$	$R_5$	KB
false false : false	false false : true	false false : false	false false : false	false false : false	false false : false	false true : false	true true : true	$true$ $true$ $\vdots$ $true$	true false : false	$true$ $true$ $\vdots$ $true$	false false : true	false false : false
false false false	true true true	false false false	false false false	false false false	false true true	true false true	true true true	true true true	true true true	true true true	true true true	$\frac{true}{true}$
false : true	true : true	false : true	false : true	true : true	false : true	false : true	true : false	false : true	false : true	true : false	true : true	false : false

#### Inference by (depth-first) enumeration

```
function TT-ENTAILS?(KB,\alpha) returns true or false
  inputs: KB, the knowledge base, a sentence in propositional logic
                  \alpha, the query, a sentence in propositional logic
  symbols \leftarrow a list of the proposition symbols in KB and \alpha
  return TT-CHECK-ALL(KB,\alpha,symbols,\{\})
function TT-CHECK-ALL(KB,\alpha,symbols,model) returns true or false
  if EMPTY?(symbols) then
    if PL-TRUE?(KB,model) then return PL-TRUE?(\alpha,model)
    else return true
                           // when KB is false, always return true
  else do
                          sound and complete
    P \leftarrow \text{FIRST}(symbols) Time complexity O(2^n), space complexity O(n)
    rest \leftarrow REST(symbols)
    return (TT-CHECK-ALL(KB,\alpha,rest,model \cup \{P = true\})
         and TT-CHECK-ALL(KB,\alpha,rest,model \cup \{P = false\}))
```

#### Quiz 01: Model-checking approach

Given a KB containing the following rules and facts

R<sub>1</sub>: IF hot AND smoky THEN fire

R<sub>2</sub>: IF alarm\_beeps THEN smoky

R<sub>3</sub>: IF fire THEN sprinklers\_on

F₁: alarm\_beeps

F<sub>2</sub>: hot

- Represent the KB in propositional logic with given symbols
  - H = hot, S = smoky, F = fire, A = alarms\_beeps, R = sprinklers\_on
- Answer the question "Sprinklers\_on?" by using the modelchecking approach.

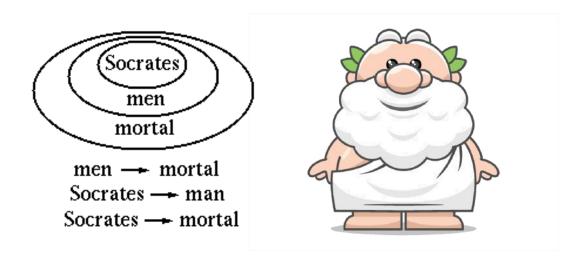


## Propositional theorem proving

- Proof by Resolution
- Forward and Backward Chaining

#### Inference rules approach

- Theorem proving: Apply rules of inference directly to the sentences in KB to construct a proof of the desired sentence without consulting models
- More efficient than model checking when the number of models is large, yet the length of the proof is short



#### Logical equivalence

• Two sentences,  $\alpha$  and  $\beta$ , are logically equivalent if they are true in the same set of models.

$$\alpha \equiv \beta \text{ iff } \alpha \models \beta \text{ and } \beta \models \alpha$$

```
(\alpha \wedge \beta) \equiv (\beta \wedge \alpha) commutativity of \wedge
           (\alpha \vee \beta) \equiv (\beta \vee \alpha) commutativity of \vee
((\alpha \wedge \beta) \wedge \gamma) \equiv (\alpha \wedge (\beta \wedge \gamma)) associativity of \wedge
((\alpha \vee \beta) \vee \gamma) \equiv (\alpha \vee (\beta \vee \gamma)) associativity of \vee
            \neg(\neg\alpha) \equiv \alpha double-negation elimination
       (\alpha \Rightarrow \beta) \equiv (\neg \beta \Rightarrow \neg \alpha) contraposition
      (\alpha \Rightarrow \beta) \equiv (\neg \alpha \lor \beta) implication elimination
      (\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)) biconditional elimination
       \neg(\alpha \land \beta) \equiv (\neg \alpha \lor \neg \beta) De Morgan
        \neg(\alpha \lor \beta) \equiv (\neg \alpha \land \neg \beta) De Morgan
(\alpha \wedge (\beta \vee \gamma)) \equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) distributivity of \wedge over \vee
(\alpha \vee (\beta \wedge \gamma)) \equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) distributivity of \vee over \wedge
```

#### Validity

- A sentence is valid if it is true in all models.
  - E.g.,  $P \vee \neg P$ ,  $P \Rightarrow \neg P$ ,  $(P \wedge (P \Rightarrow Q)) \Rightarrow Q$
- Valid sentences are also known as tautologies.
- Validity is connected to inference via the Deduction Theorem

$$\alpha \models \beta \text{ if } f \ \alpha \Rightarrow \beta \text{ is valid}$$

#### Satisfiability

- A sentence is satisfiable if it is true in some model.
  - E.g., *P* ∨ *Q*, *P*
- A sentence is unsatisfiable if it is true in no models.
  - E.g.,  $P \land \neg P$
- Satisfiability is connected to inference via the following  $\alpha \models \beta \ iff \ \alpha \land \neg \beta \ is \ unsatisfiable$ 
  - → Refutation or proof by contradiction
- The SAT problem determines the satisfiability of sentences in propositional logic (NP-complete)
  - E.g., in CSPs, the constraints are satisfiable by some assignment.

#### Quiz 02: Validity and Satisfiability

 Check the validity and satisfiability of the below sentences using the truth table

- 1.  $A \lor B \Rightarrow A \land C$
- 2.  $A \wedge B \Rightarrow A \vee C$
- 3.  $(A \lor B) \land (\neg B \lor C) \Rightarrow A \lor C$
- 4.  $(A \lor \neg B) \Rightarrow A \land B$

#### Inference and Proofs

- Proof: A chain of conclusions leads to the desired goal
- Example sound rules of inference

$$\begin{array}{c} \alpha \Rightarrow \beta \\ \alpha \\ \hline \vdots \beta \end{array}$$

$$\begin{array}{ccc} \Rightarrow \beta & & \alpha \Rightarrow \beta \\ \hline & & -\beta \\ \hline \beta & & \vdots \neg \alpha \end{array}$$

$$\begin{array}{ccc} \alpha & & \alpha \wedge \beta \\ \\ \beta & & \\ \vdots & \alpha \wedge \beta & & \\ \end{array}$$

$$\begin{array}{ccc} & & \\ & \vdots & \alpha \end{array}$$

Modus Ponens

Modus Tollens

**AND-Introduction** 

**AND-Elimination** 

#### Inference rules: An example

KB
$P \wedge Q$
$P \Rightarrow R$
$Q \wedge R \Rightarrow S$

**S**?

No.	Sentences	Explanation
1	$P \wedge Q$	From KB
2	$P \Rightarrow R$	From KB
3	$Q \wedge R \Rightarrow S$	From KB
4	P	1 And-Elim
5	R	4,2 Modus Ponens
6	Q	1 And-Elim
7	$Q \wedge R$	5,6 And-Intro
8	S	3,7 Modus Ponens

# Inference rules in Wumpus world

$$R_1$$
:  $\neg P_{1.1}$ 

$$R_2: B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$$

$$R_3: B_{2,1} \Leftrightarrow (P_{1,1} \vee P_{2,2} \vee P_{3,1})$$

$$R_4$$
:  $\neg B_{1,1}$ 

$$R_5$$
:  $B_{2,1}$ 

Proof:  $\neg P_{1,2}$ 

- Bi-conditional elimination to  $R_2$ :  $R_6$ :  $\left(B_{1,1} \Rightarrow \left(P_{1,2} \vee P_{2,1}\right)\right) \wedge \left(\left(P_{1,2} \vee P_{2,1}\right) \Rightarrow B_{1,1}\right)$
- And-Elimination to  $R_6$ :  $R_7$ :  $(P_{1,2} \lor P_{2,1}) \Rightarrow B_{1,1}$
- Logical equivalence for contrapositives:  $R_8: \neg B_{1,1} \Rightarrow \neg (P_{1,2} \lor P_{2,1})$
- Modus Ponens with  $R_8$  and the percept  $R_4$ :  $R_9$ :  $\neg(P_{1,2} \lor P_{2,1})$
- De Morgan's rule:  $R_{10}$ :  $\neg P_{1,2} \land \neg P_{2,1}$

# Proving by search

- Search algorithms can be applied to find a sequence of steps that constitutes a proof.
  - INITIAL STATE: the initial knowledge base
  - ACTIONS: apply all inference rules to all the sentences that match the top half of the inference rule
  - RESULT: add the sentence in the bottom half of the inference rule
  - GOAL: a state that contains the sentence need to be proved
- The proof can ignore irrelevant propositions, no matter how many of them there are → more efficient
  - E.g., in the Wumpus world,  $B_{2,1}$ ,  $P_{1,1}$ ,  $P_{2,2}$  and  $P_{3,1}$  are not mentioned.

#### Monotonicity

 The set of entailed sentences only increases as information is added to the knowledge base.

if 
$$KB \models \alpha$$
 then  $KB \land \beta \models \alpha$ 

• Additional conclusions can be drawn without invalidating any conclusion  $\alpha$  already inferred.

## Proof by Resolution

- Proof by Inference Rules: sound but not complete
  - If the rules are inadequate, then the goal is not reachable.
- Resolution: sound and complete, a single inference rule
  - A **complete** inference algorithm when coupled with any complete search algorithm  $l_1 \vee \cdots \vee l_k$
  - Unit resolution inference rule

where  $l_i$  and m are complementary literals

$$\frac{m}{l_1 \vee \dots \vee l_{i-1} \vee l_{i+1} \vee \dots \vee l_k}$$

Full resolution rule

$$l_1 \lor \cdots \lor l_k$$
 $m_1 \lor \cdots \lor m_n$ 

$$l_1 \vee \cdots \vee l_{i-1} \vee l_{i+1} \vee \cdots \vee l_k \vee m_1 \vee \cdots \vee m_{j-1} \vee m_{j+1} \vee \cdots \vee m_n$$

where  $l_i$  and  $m_i$  are complementary literals

# Inference rules in Wumpus world

$$R_1$$
:  $\neg P_{1,1}$ 

$$R_2: B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$$

$$R_3: B_{2,1} \Leftrightarrow (P_{1,1} \vee P_{2,2} \vee P_{3,1})$$

$$R_4$$
:  $\neg B_{1.1}$ 

$$R_5$$
:  $B_{2.1}$ 

$$R_6: \left(B_{1,1} \Rightarrow \left(P_{1,2} \vee P_{2,1}\right)\right) \wedge \left(\left(P_{1,2} \vee P_{2,1}\right) \Rightarrow B_{1,1}\right)$$

$$R_7$$
:  $\neg P_{1,2} \land \neg P_{2,1} \Rightarrow B_{1,1}$ 

$$R_8: \neg B_{1,1} \Rightarrow \neg (P_{1,2} \vee P_{2,1})$$

$$R_9 : \neg (P_{1,2} \vee P_{2,1})$$

$$R_{10} : \neg P_{1,2} \land \neg P_{2,1}$$

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2	2,2 <b>P</b> ?	3,2	4,2
OK			
1,1	2,1 A	3,1 P?	4,1
V	B		
OK	OK		

# Inference rules in Wumpus world

$$R_1$$
:  $\neg P_{1,1}$ 

...

$$R_{11}$$
:  $\neg B_{1,2}$ 

$$R_{12}: B_{1,2} \Leftrightarrow (P_{1,1} \vee P_{2,2} \vee P_{1,3})$$

$$R_{13}$$
:  $\neg P_{2,2}$ 

$$R_{14}$$
:  $\neg P_{1.3}$ 

$$R_{15}$$
:  $P_{1,1} \vee P_{2,2} \vee P_{3,1}$ 

$$R_{16}$$
:  $P_{1,1} \vee P_{3,1}$ 

$$R_{17}$$
:  $P_{3,1}$ 

1,4	2,4	3,4	4,4
1,3 W!	2,3	3,3	4,3
1,2A S OK	2,2 OK	3,2	4,2
1,1 V OK	2,1 B V OK	3,1 P!	4,1

 $\neg P_{2,2}$  resolves with  $P_{2,2}$ 

 $\neg P_{1,1}$  resolves with  $P_{1,1}$ 

## Proof by Resolution

- Factoring: the resulting clause should contain only one copy of each literal.
  - E.g., resolving  $(A \lor B)$  with  $(A \lor \neg B)$  obtains  $(A \lor A) \to \text{reduced to } A$
- For any pair of sentences,  $\alpha$  and  $\beta$ , in propositional logic, a resolution-based theorem prover can decide whether  $\alpha \models \beta$ .

# Conjunctive Normal Form (CNF)

- Resolution applies only to clauses, i.e., disjunctions of literals
  - → Convert all sentences in KB into clauses (CNF form)
- For example, convert  $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$  into CNF

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

→ A conjunction of 3 clauses

#### Conversion to CNF

- 1. Eliminate  $\Leftrightarrow$ :  $\alpha \Leftrightarrow \beta \equiv (\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$
- 2. Eliminate  $\Rightarrow$ :  $\alpha \Rightarrow \beta \equiv \neg \alpha \lor \beta$
- 3. The operator ¬ appears only in literals: "move ¬ inwards"

 $\neg \neg \alpha \equiv \alpha$  (double-negation elimination)

 $\neg(\alpha \land \beta) \equiv \neg\alpha \lor \neg\beta$  (De Morgan)

 $\neg(\alpha \lor \beta) \equiv \neg\alpha \land \neg\beta$  (De Morgan)

4. Apply the distributivity law to distribute  $\vee$  over  $\wedge$ 

$$(\alpha \land \beta) \lor \gamma \equiv (\alpha \lor \gamma) \land (\beta \lor \gamma)$$

#### Quiz 03: Conversion to CNF

Convert the following sentences into CNF

1. 
$$(A \land B) \Rightarrow (C \Rightarrow D)$$

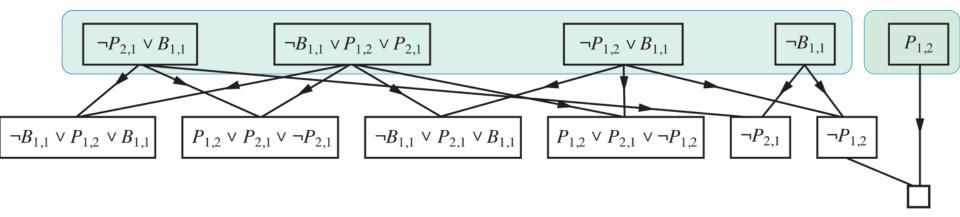
2. 
$$P \lor Q \Leftrightarrow R \land \neg Q \Rightarrow P$$

## The resolution algorithm

• Proof by contradiction (resolution refutation): To show that  $KB \models \alpha$ , prove  $KB \land \neg \alpha$  is unsatisfiable

```
function PL-RESOLUTION(KB,\alpha) returns true or false
  inputs: KB, the knowledge base, a sentence in propositional logic
                   \alpha, the query, a sentence in propositional logic
  clauses ← the set of clauses in the CNF representation of KB \land \neg \alpha
  new \leftarrow \{ \}
  loop do
    for each pair of clauses C_i, C_i in clauses do
       resolvents \leftarrow PL-RESOLVE(C_i, C_i)
       if resolvents contains the empty clause then return true
       new \leftarrow new \cup resolvents
    if new \subseteq clauses then return false
     clauses \leftarrow clauses \cup new
```

## The resolution algorithm



- Many resolution steps are pointless.
- Clauses with two complementary literals can be discarded.
  - E.g.,  $B_{1,1} \vee \neg B_{1,1} \vee P_{2,1} \equiv True \vee P_{2,1} \equiv True$

# Quiz 04: The resolution algorithm

- Given the following hypotheses
  - If it rains, Joe brings his umbrella.
  - If Joe brings his umbrella, Joe does not get wet.
  - If it does not rain, Joe does not get wet.
- Prove that Joe does not get wet.

# Quiz 04: The resolution algorithm

The KB contains facts and hypotheses

# KB $R \Rightarrow U$ $U \Rightarrow \neg W$ $\neg R \Rightarrow \neg W$

- Check if the sentence
- $\neg W$  is entailed by KB?

#### Horn clauses and Definite clauses

- Definite clause: a disjunction of literals of which exactly one is positive.
  - E.g.,  $\neg P \lor \neg Q \lor R$  is a definite clause, whereas  $\neg P \lor Q \lor R$  is not.
- Horn clause: a disjunction of literals of which at most one is positive.
  - All definite clauses are Horn clauses
- Goal clause: clauses with no positive literals
- Horn clauses are closed under resolution
  - Resolving two Horn clauses will get back a Horn clause.

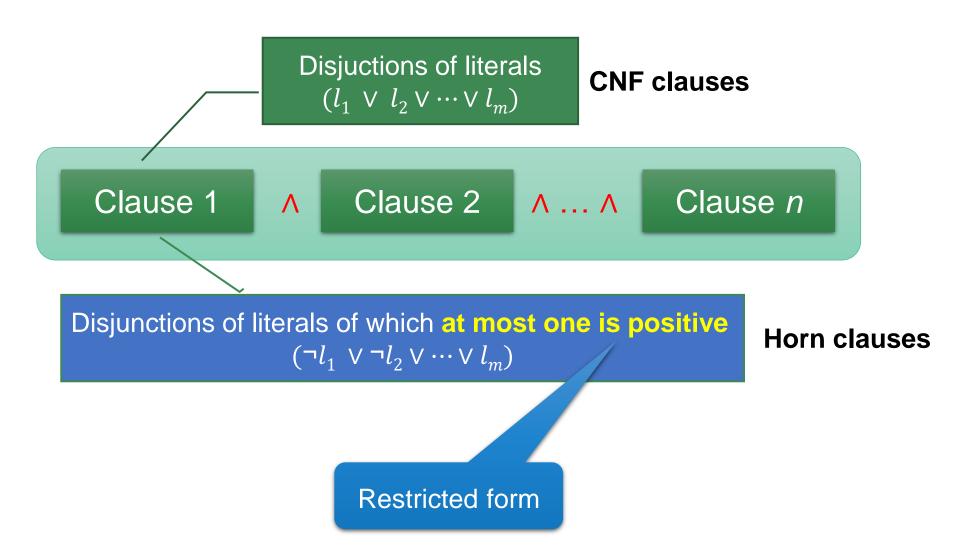
# Backus normal form (BNF)

```
CNFSentence \rightarrow Clause_1 \wedge \cdots \wedge Clause_n
                  Clause \rightarrow Literal_1 \vee \cdots \vee Literal_m
                  Literal \rightarrow Symbol \mid \neg Symbol
                 Symbol \rightarrow P \mid Q \mid R \mid \dots
   HornClauseForm \rightarrow DefiniteClauseForm \mid GoalClauseForm
DefiniteClauseForm \rightarrow (Symbol_1 \land \cdots \land Symbol_1) \Rightarrow Symbol_1
    GoalClauseForm \rightarrow (Symbol_1 \land \cdots \land Symbol_l) \Rightarrow False
```

#### KB of definite clauses

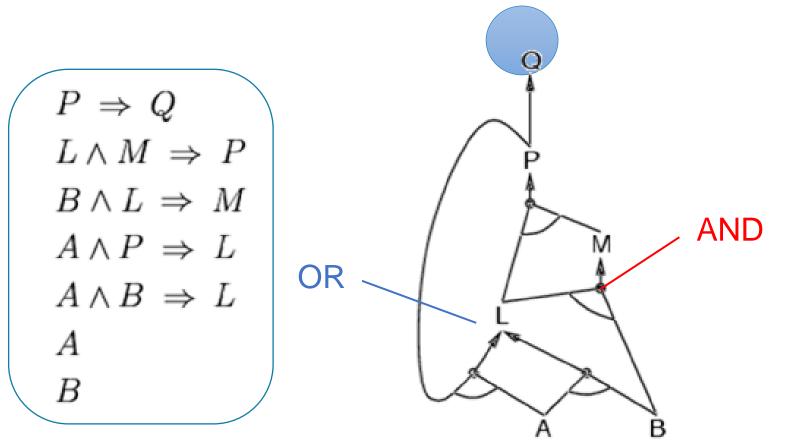
- KB containing only definite clauses are interesting.
- Every definite clause can be written as an implication.
  - Premise (body) is a conjunction of positive literals and Conclusion (head) is a single positive literal (fact) → easier to understand
  - E.g.,  $\neg P \lor \neg Q \lor R \equiv (P \land Q) \Rightarrow R$
- Inference can be done with forward-chaining and backwardchaining algorithms
  - This type of inference is the basis for logic programming.
- Deciding entailment can be done in linear time.

#### KB: Horn clauses vs. CNF clauses



## Forward chaining

 Key idea: Fire any rule whose premises are satisfied in the KB, add its conclusion to the KB, until the query is found.



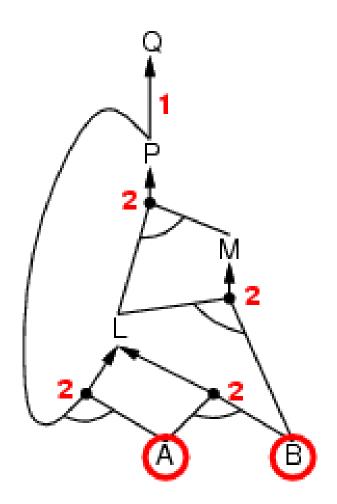
# The forward chaining algorithm

```
function PL-FC-ENTAILS?(KB, q) returns true or false
  inputs: KB, the knowledge base, a set of propositional definite clauses
           q, the query, a proposition symbol
  count \leftarrow a table, where count[c] is the number of symbols in c's premise
  inferred \leftarrow a table, where inferred[s] is initially false for all symbols
  agenda \leftarrow a queue of symbols, initially symbols known to be true in KB
  while agenda is not empty do
    p \leftarrow POP(agenda)
                                         Sound and complete
    if p = q then return true
    if inferred[p] = false then
      inferred[p] \leftarrow true
      for each clause c in KB where p is in c.PREMISE do
         decrement count[c]
```

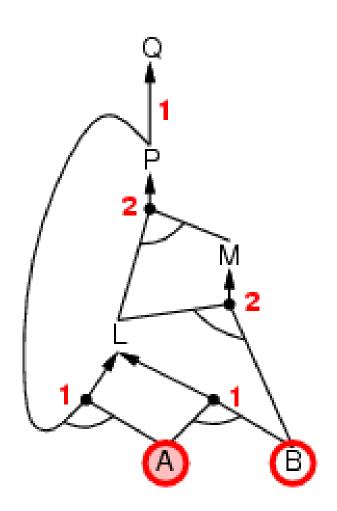
return false

**if** count[c] = 0 **then** add c.CONCLUSION to agenda

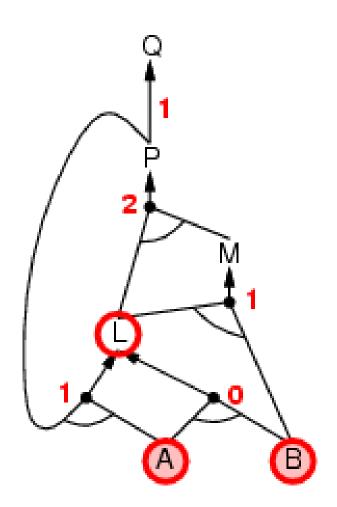
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



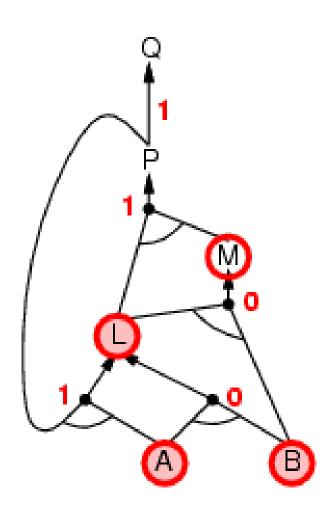
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



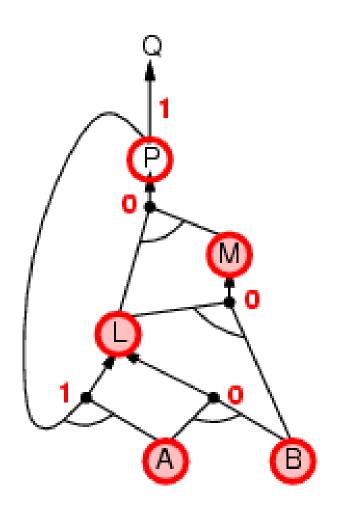
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



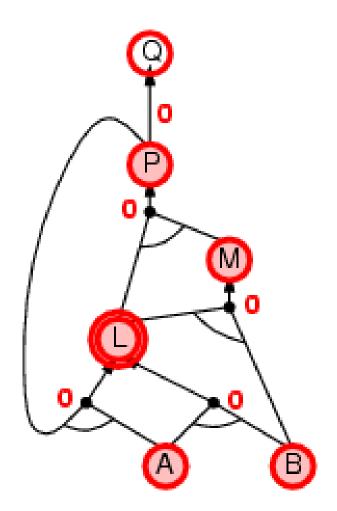
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



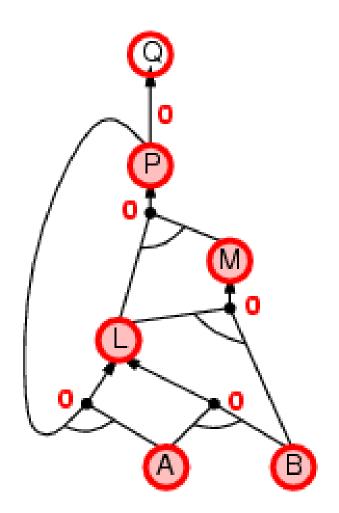
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



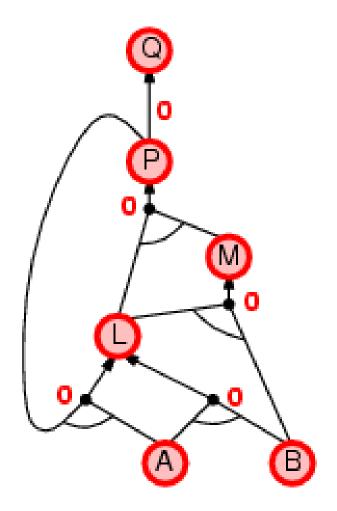
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



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



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



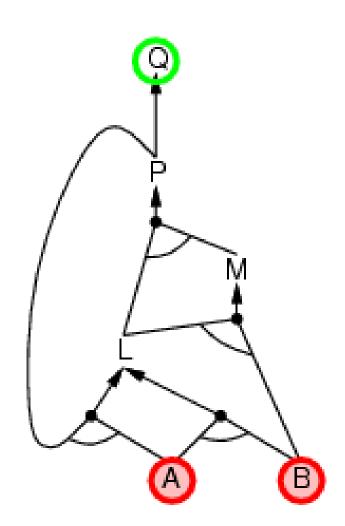
KB
$A \wedge B \Rightarrow C$
$C \wedge D \Rightarrow E$
$C \wedge F \Rightarrow G$
$\boldsymbol{A}$
B
D
<b>E</b> ?

No.	Sentences	Explanation
1	$A \wedge B \Rightarrow C$	From KB
2	$C \wedge D \Rightarrow E$	From KB
3	$C \wedge F \Rightarrow G$	From KB
4	$\boldsymbol{A}$	From KB
5	B	From KB
6	D	From KB
7	С	1, 4 and 5
8	<b>E</b>	2, 6, and 7

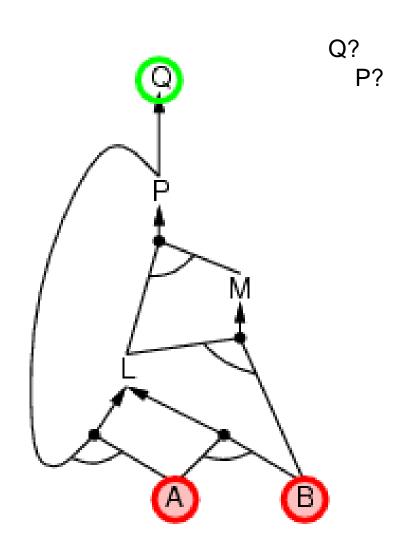
#### Backward chaining

- Key idea: Work backwards from the query q
  - Check if q is known already, or
  - Recursively prove by BC all premises of some rule concluding q
- Avoid loops: A new subgoal is already on the goal stack?
- Avoid repeated work: A new subgoal has already been proved true, or has already failed?

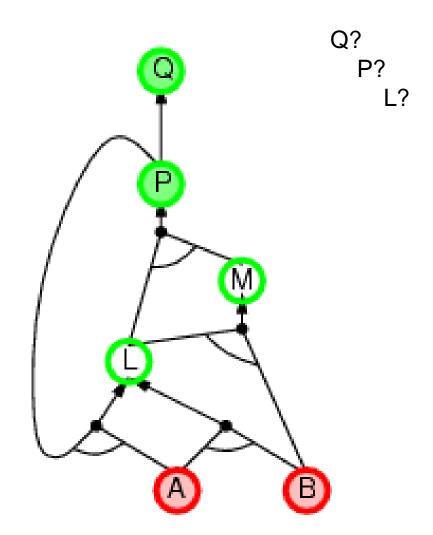
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 

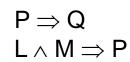


$$P \Rightarrow Q$$
 $L \wedge M \Rightarrow P$ 
 $B \wedge L \Rightarrow M$ 
 $A \wedge P \Rightarrow L$ 
 $A \wedge B \Rightarrow L$ 
 $A$ 

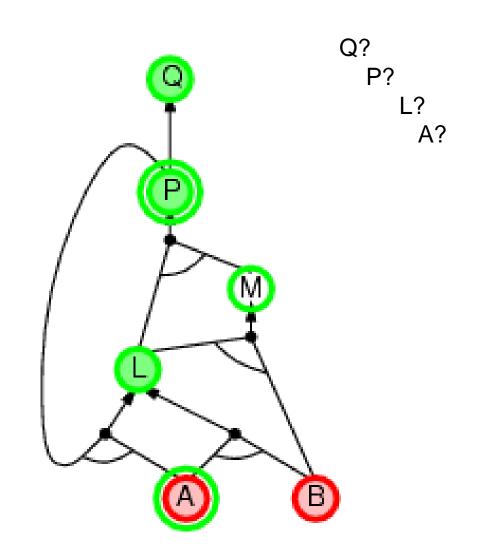


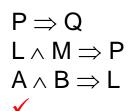
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



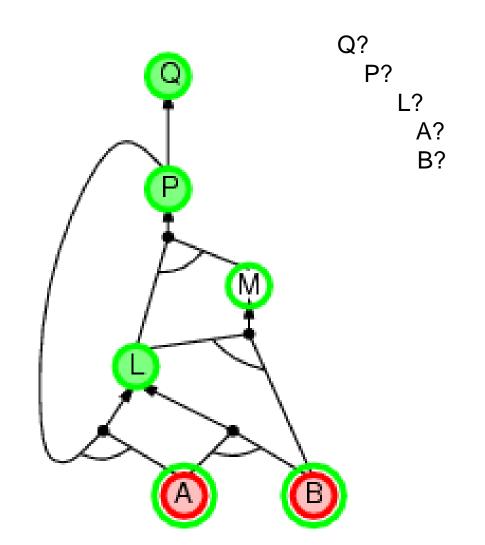


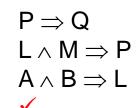
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



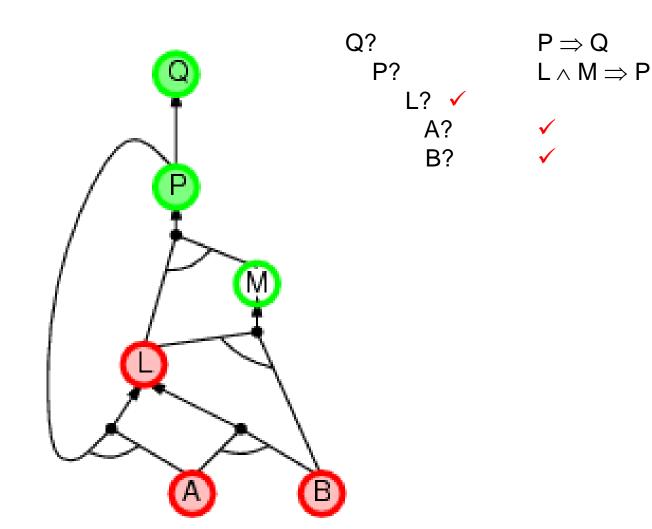


$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 

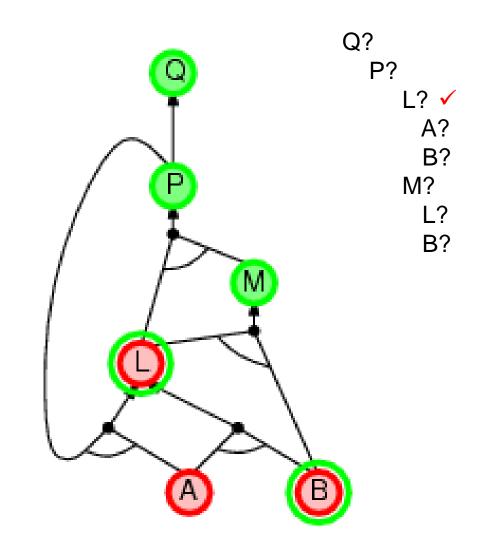


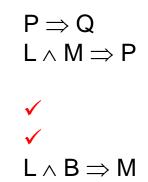


$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 

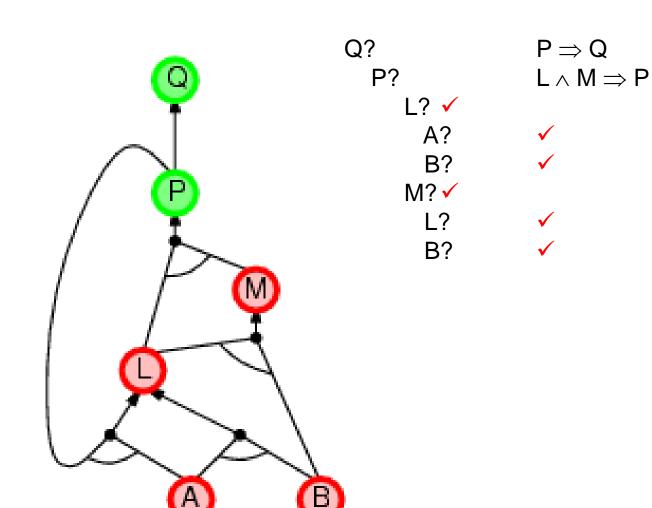


$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 

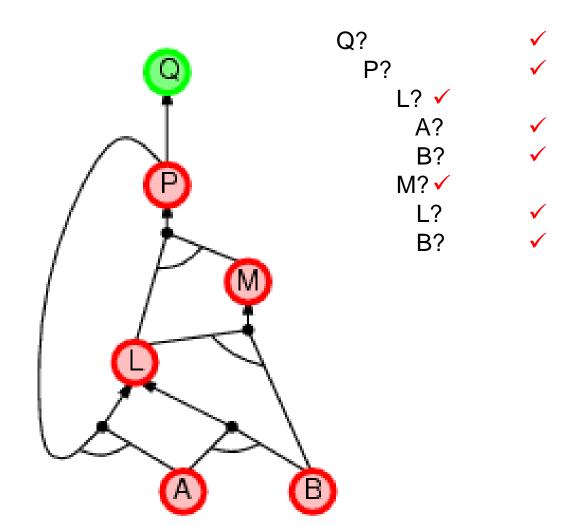




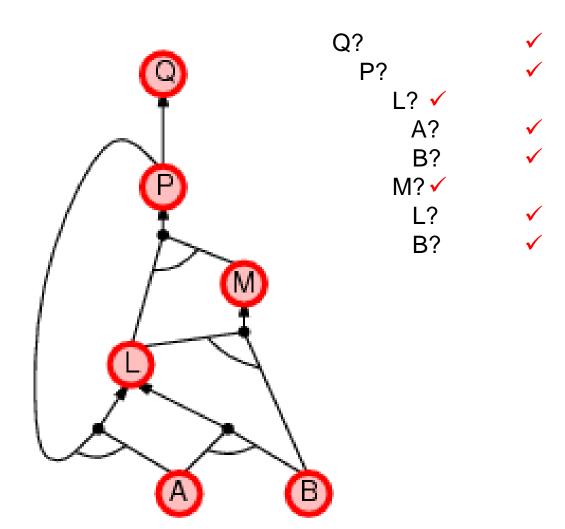
$$P \Rightarrow Q$$
 $L \land M \Rightarrow P$ 
 $B \land L \Rightarrow M$ 
 $A \land P \Rightarrow L$ 
 $A \land B \Rightarrow L$ 
 $A$ 



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



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



#### **KB**

$$A \wedge B \Rightarrow C$$

$$C \wedge D \Rightarrow E$$

$$C \wedge F \Rightarrow G$$

 $\boldsymbol{A}$ 

B

D

E?

- E?
  - C?
    - A?
    - B?
  - D?
- A, B and D are given → All needed rules are satisfied → The goal is proven.

 $A \wedge B \Rightarrow C$ 

#### Forward vs. Backward chaining

- Forward chaining: data-driven, automatic, unconscious processing
  - E.g., object recognition, routine decisions
  - May do lots of work that is irrelevant to the goal
- Backward chaining: goal-driven, good for problem-solving
  - E.g., Where are my keys? How do I get into a PhD program?
  - Complexity can be much less than linear in size of KB

#### Quiz 05: Forward vs. Backward chaining

Given a KB containing the following rules and facts

R<sub>1</sub>: IF hot AND smoky THEN fire

R<sub>2</sub>: IF alarm\_beeps THEN smoky

R<sub>3</sub>: IF fire THEN sprinklers\_on

F₁: alarm\_beeps

 $F_2$ : hot

- Represent the KB in propositional logic with given symbols
  - H = hot, S = smoky, F = fire, A = alarms\_beeps, R = sprinklers\_on
- Answer the question "Sprinklers\_on?" by using the forward chaining and backward chaining approaches



# Effective model checking

- A complete backtracking algorithm
- Local search algorithms

#### Efficient propositional inference

- The SAT problem (checking satisfiability)
  - Testing entailment,  $\alpha \models \beta$ ? = testing unsatisfiability of  $\alpha \land \neg \beta$

- Two families of efficient algorithms for general propositional inference based on model checking
  - 1. Complete backtracking search algorithms
    - DPLL algorithm (Davis, Putnam, Logemann, Loveland)
  - 2. Incomplete local search algorithms (hill-climbing)
    - WalkSAT algorithm

#### The DPLL algorithm

- Often called the Davis-Putnam algorithm (1960)
- Determine whether an input propositional logic sentence (in CNF) is satisfiable.
  - A recursive, depth-first enumeration of possible models.
- Improvements over truth table enumeration
  - 1. Early termination
  - 2. Pure symbol heuristic
  - 3. Unit clause heuristic

#### Improvements in DPLL

- Early termination: A clause is true if any literal is true, and a sentence is false if any clause is false.
  - Avoid examination of entire subtrees in the search space
  - E.g.,  $(A \lor B) \land (A \lor C)$  is true if A is true, regardless B and C
- Pure symbol heuristic: A pure symbol always appears with the same "sign" in all clauses.
  - E.g.,  $(A \lor \neg B)$ ,  $(\neg B \lor \neg C)$ ,  $(A \lor C)$ , A and B are pure, C is impure.
  - Make a pure symbol true → Doing so never make a clause false
- Unit clause heuristic: there is only one literal in the clause and thus this literal must be true
  - **Unit propagation:** if the model contains B = true then  $(\neg B \lor \neg C)$  simplifies to a unit clause  $\neg C \to C$  must be false (so that  $\neg C$  is true)  $\to A$  must be true (so that  $A \lor C$  is true)

## The DPLL procedure

function DPLL-SATISFIABLE?(s) returns true or false
 inputs: s, a sentence in propositional logic
 clauses ← the set of clauses in the CNF representation of s
 symbols ← a list of the proposition symbols in s
 return DPLL(clauses, symbols,{})

```
function DPLL(clauses, symbols, model) returns true or false
  if every clause in clauses is true in model then return true
                                                                           1. Early
  if some clause in clauses is false in model then return false -
                                                                           Termination
  P, value \leftarrow FIND-PURE-SYMBOL (symbols, clauses, model)
  if P is non-null then return DPLL(clauses, symbols – P, model \cup {P=value})
  P, value \leftarrow FIND-UNIT-CLAUSE (clauses, model)
  if P is non-null then return \boxed{\text{DPLL}(clauses, symbols - P, model <math>\cup \{P=value\})}
  P \leftarrow \text{FIRST}(symbols); rest \leftarrow \text{REST}(symbols)
  return DPLL(clauses, rest, model \cup {P=true}) or
     DPLL(clauses, rest, model \cup \{P=false\}))
```

# The Davis-Putnam procedure

```
function DP(\Delta)
   for \phi in vocabulary (\Delta) do
      var \Delta' \leftarrow \{ \};
      for \Phi_1 in \Delta for \Phi_2 in \Delta such that \varphi \in \Phi_1 \neg \varphi \in \Phi_2 do
             \operatorname{var} \Phi' \leftarrow \Phi_1 - \{ \varphi \} \cup \Phi_2 - \{ \neg \varphi \};
             if not tautology(\Phi') then \Delta' \leftarrow \Delta' \cup (\Phi');
       \Delta \leftarrow \Delta - \{\Phi \in \Delta \mid \varphi \in \Phi \text{ or } \neg \varphi \in \Phi\} \cup \Delta';
   return {if { } \in \Delta then unsatisfiable else satisfiable};
function tautology(\Phi)
    \varphi \in \Phi and \neg \varphi \in \Phi
```

#### DPLL procedure vs. DP procedure

- DP can cause a quadratic expansion every time it is applied.
  - This can easily exhaust space on large problems.
- DPLL attacks the problem by sequentially solving smaller problems.
  - Basic idea: Choose a literal. Assume true, simplify clause set, and try to show satisfiable. Repeat for the negation of the literal.
  - Good because we do not cross multiply the clause set

# DPLL procedure vs. DP procedure

Problem	Tautology	DP	DPLL
Prime	30.00	0.00	0.00
Prime4	0.02	0.06	0.04
Prime9	18.94	2.98	0.51
Prime10	11.40	3.03	0.96
Prime11	28.11	2.98	0.51
Prime16	> 1 hour	*	9.15
Prime17	> 1 hour	*	3.87
Mkadder32	>> 1 hour	6.50	7.34
Mkadder42	>> 1 hour	22.95	46.86
Mkadder52	>> 1 hour	44.83	170.98
Mkadder53	>> 1 hour	38.27	250.16
Mkadder63	>> 1 hour	*	1186.4
Mkadder73	>> 1 hour	*	3759.9

Reference: http://logic.stanford.edu/classes/cs157/2011/lectures/lecture04.pdf

#### The WalkSAT algorithm

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

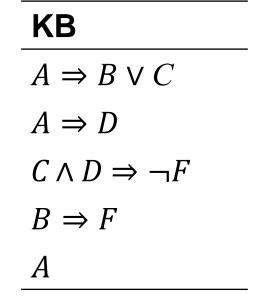
**function** WALKSAT(clauses, p, max\_flips) **returns** a satisfying model or failure

#### The WalkSAT algorithm

- The algorithm returns a model → satisfiable
- The algorithm returns false → unsatisfiable OR more time is needed for searching
- WalkSAT cannot always detect unsatisfiability
  - It is most useful when a solution is expected to exist.
- For example,
  - An agent cannot *reliably* use WALKSAT to prove that a square is safe in the Wumpus world.
  - Instead, it can say, "I thought about it for an hour and couldn't come up with a possible world in which the square isn't safe."

#### Quiz 06: DPLL and DP

Given a KB as shown aside



- Using either DPLL or DP to check whether KB entails each of the following sentences
  - C
  - $B \Rightarrow \neg C$

... the end.