

# *Introduction to Logical Agents (Propositional Logic)*

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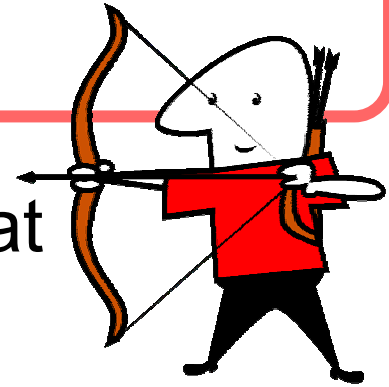
Thanks to Professor Andrew W. Moore (Carnegie Mellon University) <http://www.cs.cmu.edu/~awm/tutorials>  
Also: Artificial Intelligence: A Modern Approach, 2<sup>nd</sup> Ed., Russel & Norvig



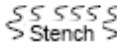



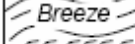
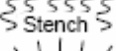


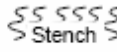



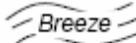
# The “Wumpus” World



A very simple computer game (somewhat similar to minesweeper...)



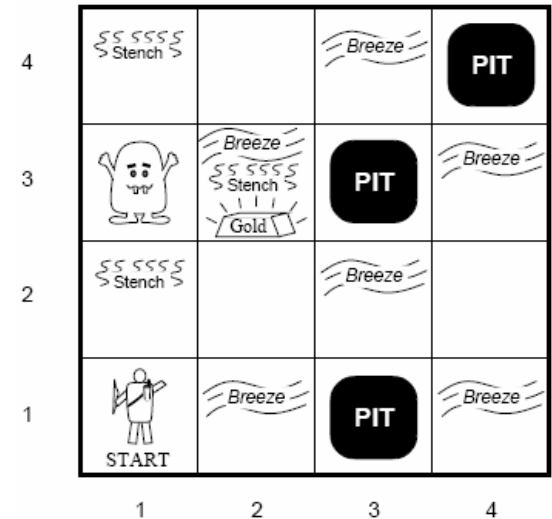
- An agent is exploring,
- Avoid deadly wumpus and pits,
- Find the gold!
- Given: Certain laws of the world + agent's observations

4	 Stench		 Breeze	<b>PIT</b>
3	 	 Breeze  Stench 	<b>PIT</b>	 Breeze
2	 Stench		 Breeze	
1	 START	 Breeze	<b>PIT</b>	 Breeze
	1	2	3	4



# The “Wumpus” World – PEAS

- *Performance measure*
  - gold: +1000, death: -1000, -1 per step, -10 for using the arrow
- *Environment*
  - Squares adjacent to wumpus are smelly
  - Squares adjacent to pit are breezy
  - Glitter i gold is in the same square
  - Shooting kills wumpus if you are facing it
  - Shooting uses up the only arrow
  - Grabbing picks up gold if in same square
  - Releasing drops the gold in same square
- *Actuators*
  - Left turn, Right turn, Forward, Grab, Release, Shoot
- *Sensors*
  - Breeze, Glitter, Smell





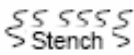
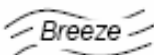

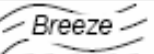
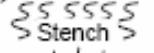
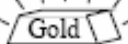
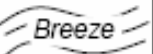
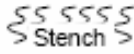
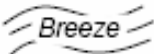



# Wumpus World - characterization

- As viewed by the agent...
- Observable ?? No -- only local perception
- Deterministic ?? Yes -- outcomes exactly specified.
- Episodic ?? No -- sequential at the level of actions.
- Static ?? Yes -- wumpus and pits do not move
- Discrete ?? Yes -- finite number of states
- Single-Agent ?? Yes -- wumpus is essentially a natural feature




# Exploring a wumpus world...

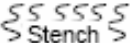



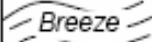
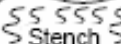


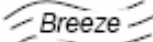
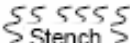
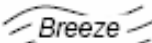

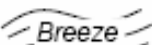

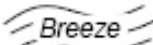
What we do not know...

4				<b>PIT</b>
3		  	<b>PIT</b>	
2				
1	 START		<b>PIT</b>	
	1	2	3	4

Here is what we know:

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



4	 Stench		 Breeze	 PIT
3		 Breeze  Stench  Gold	 PIT	 Breeze
2	 Stench		 Breeze	
1	 START	 Breeze	 PIT	 Breeze
	1	2	3	4

6



# Exploring a wumpus world...

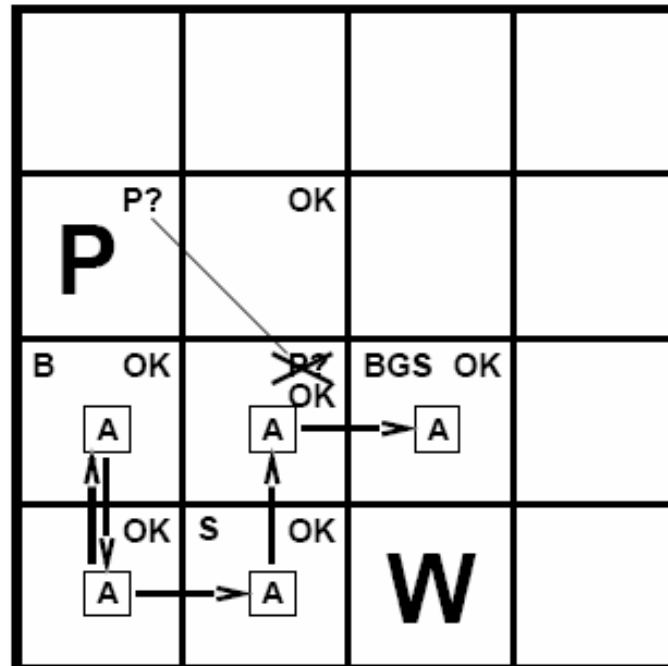
- Now another (unknown) world... How would you explore?
- Let us try!

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
OK			
1,2	2,2	3,2	4,2
OK	OK		
<div>1,1<div>A</div></div>	2,1	3,1	4,1



# Exploring a wumpus world...

- Our final knowledge about this world...

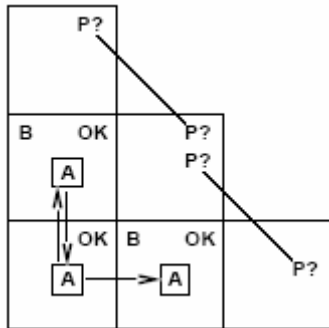




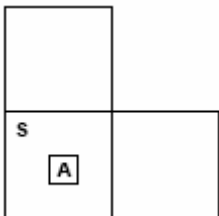


# Exploring a wumpus world...

- Some tight spots



- Breeze in 1,2 and 2,1 !! → No safe actions
- However: Assuming pits uniformly distributed →
- 2,2 has pit with prob. 0.86 vs 0.31



- Stench in 1,1 → Cannot move!!
- Can use a strategy of *coercion*:
- Shoot straight ahead
- Wumpus was there → dead → safe
- Wumpus was not there → safe



## So what is Logic anyways?

- Logics (!) are formal languages for representing information,
- such that conclusions can be drawn.
- Syntax determines the *structure* of sentences in the language,
- Semantics determines the *meaning* of these sentences.
- i.e., determines the truth of a sentence in a model (also called “possible world”)



## So what is Logic anyways?

- Example: The language of arithmetic:
- $x+2 \geq y$  is a sentence;  $x^2+y >$  is not a sentence
- $x+2 \geq y$  is true iff the number  $x+2$  is no less than the number  $y$
- $x+2 \geq y$  is *true* in a world where  $x=7$ ;  $y=1$
- $x+2 \geq y$  is *false* in a world where  $x=0$ ;  $y=6$



## So what is Logic anyways?

- *Declarative* versus *Procedural* system building...
- Logics are declarative languages.
- *Propositional Logic*: A very simple logic
- Also called: *Binary Logic*



# Propositional Logic Sentences

- $a \wedge b$
- Sunny  $\vee$  Cloudy
- $\sim(\text{AmTired} \wedge \text{AmEnergetic})$
- $\text{LectureBoring} \Rightarrow \text{InstructorFired}$
- $\sim(\text{LectureBoring} \Rightarrow \text{InstructorFired})$

*Syntax involves Propositional Symbols, TRUE, FALSE, the unary “ $\sim$ ” (Not) operator and  $\wedge$ ,  $\vee$ ,  $\Rightarrow$ ,  $\Leftrightarrow$  operators.*

A more formal specification is usually needed:  
See “Backus-Naur” Form Grammar



# Propositional Logic Sentences

- $\sim$  (or  $\neg$ ) (Not) *negation*,
- $\wedge$  (And) called a *conjunction*,  
arguments called *conjuncts*
- $\vee$  (Or) called a *disjunction*,
- $\Rightarrow$  (implies) called an *implication*, or a *rule*,  
premise or antecedent  $\Rightarrow$  conclusion or consequent
- $\Leftrightarrow$  (if and only if) called a *biconditional*



# Propositional Logic Semantics

- Semantics of the language establishes the truth of all sentences
- We unary and binary connectives, we have

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



## Possible Model

- A Possible Model (also called “Possible World”)
- Given  $n$  propositional symbols, there are  $2^n$  possible models (involving every combination of TRUE and FALSE assignments to the variables)





# Possible Model

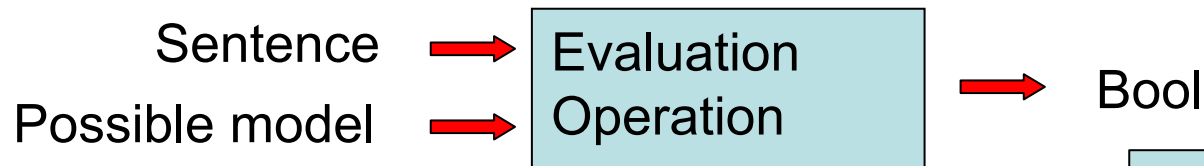
- A Possible Model (also called “Possible World”)
- Given  $n$  propositional symbols, there are  $2^n$  possible models (involving every combination of TRUE and FALSE assignments to the variables)
- Example: Suppose three propositional symbols A, B, C. There are 8 possible models.

A	B	C
FALSE	FALSE	FALSE
FALSE	FALSE	TRUE
FALSE	TRUE	FALSE
FALSE	TRUE	TRUE
TRUE	FALSE	FALSE
TRUE	FALSE	TRUE
TRUE	TRUE	FALSE
TRUE	TRUE	TRUE



# Eval(Sentence, Possible Model)

- We evaluate a *sentence* for a *possible model*.



We can express  
TRUE / FALSE by  
1 / 0

Eval("a  $\wedge$   $\sim$ (b  $\vee$  c)", 

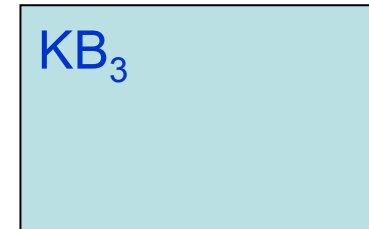
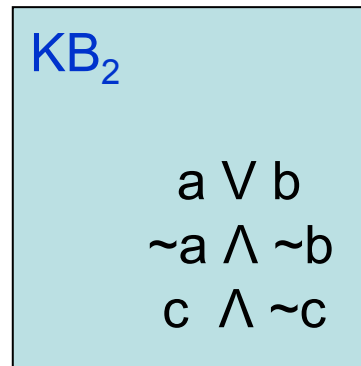
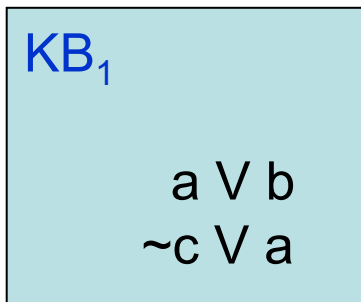
a	b	c
1	0	1

)  $\rightarrow$  0



# Knowledge Base

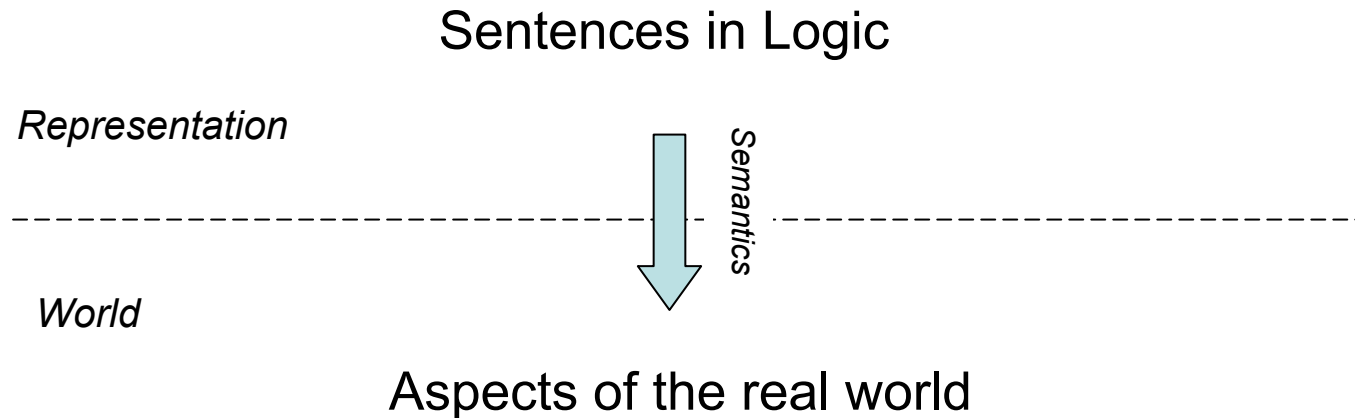
- What is it?
- It is a collection of our *knowledge about the world*,
- In logic: KB is a collection of zero or more *logic sentences*





# Knowledge Base

- Why do we care?
- Because we can *represent* the real world with this stuff !! (Umm, to some extent anyway...)





# Logical Entailment

- *Entailment* means one thing follows from another:

$$KB \models S$$

- And pronounced  
*“knowledge base KB logically entails S”*
- And means  
*“all possible models that make KB evaluate to TRUE also makes S evaluate to TRUE”*



# Logical Entailment

- $KB \models S$  pronounced  
“*knowledge base KB logically entails S*”
- And means  
“*all possible models that make KB evaluate to TRUE also makes S evaluate to TRUE*”

$KB_1$

$a \vee b$   
 $\sim c \vee a$

$a$	$b$	$c$	$KB \text{ evals to } TRUE?$	$a \wedge c$
0	0	0	0	0
0	0	1	0	0
0	1	0	1	0
0	1	1	0	0
1	0	0	1	0
1	0	1	1	1
1	1	0	1	0
1	1	1	1	1

Question:

Does  $KB \models a \wedge c$ ?



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$a$	$b$	$c$	$KB \text{ evals to } TRUE?$	$a \wedge c$
0	0	0	0	0
0	0	1	0	0
0	1	0	1	0
0	1	1	0	0
1	0	0	1	0
1	0	1	0	1
1	1	0	1	0
1	1	1	1	1

Question:

Does  $KB \models a \wedge c$ ?

Answer:

No

Because there are model in which KB is True but  $a \wedge c$  is False!



# Logical Entailment

- $KB \models S$  pronounced  
“*knowledge base KB logically entails S*”
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$KB_1$

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 $\sim c \vee a$

$a$	$b$	$c$	$KB \text{ evals to } TRUE?$	$a \vee c$
0	0	0	0	0
0	0	1	0	1
0	1	0	1	0
0	1	1	0	1
1	0	0	1	1
1	0	1	1	1
1	1	0	1	1
1	1	1	1	1

Question:

Does  $KB \models a \vee c$ ?





# Logical Entailment

- $KB \models S$  pronounced  
“*knowledge base KB logically entails S*”
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$KB_1$

$a \vee b$   
 $\sim c \vee a$

$a$	$b$	$c$	$KB \text{ evals to } TRUE?$	$a \vee c$
0	0	0	0	0
0	0	1	0	1
0	1	0	1	0
0	1	1	0	1
1	0	0	1	1
1	0	1	1	1
1	1	0	1	1
1	1	1	1	1

Question:

Does  $KB \models a \vee c$ ?

Answer:

No

Because there is a model in which KB is True but  $a \vee c$  is False!



# Logical Entailment

- $KB \models S$  pronounced  
“*knowledge base KB logically entails S*”
- And means  
“*all possible models that make KB evaluate to TRUE also makes S evaluate to TRUE*”

$KB_1$

$a \vee b$   
 $\sim c \vee a$

$a$	$b$	$c$	$KB \text{ evals to } TRUE?$	$a \vee (b \wedge \sim c) \vee (\sim b \vee c)$
0	0	0	0	0
0	0	1	0	1
0	1	0	1	1
0	1	1	0	0
1	0	0	1	1
1	0	1	1	1
1	1	0	1	1
1	1	1	1	1

Question:

Does  
 $KB \models a \vee (b \wedge \sim c) \vee (\sim b \vee c) ?$



# Logical Entailment

- $KB \models S$  pronounced  
“knowledge base  $KB$  logically entails  $S$ ”

- And means

“all possible models that make  
**TRUE** also makes  $S$  evaluate

We don't care about models that make  $KB$  false. The can make  $S$  true or false

Que

$KB_1$

$a \vee b$   
 $\sim c \vee a$

$a$	$b$	$c$	$KB$ evals to <b>TRUE?</b>	$a \vee (b \wedge \sim c)$ $\vee (\sim b \vee c)$
0	0	0	0	0
0	0	1	0	1
0	1	0	1	1
0	1	1	0	0
1	0	0	1	1
1	0	1	1	1
1	1	0	1	1
1	1	1	1	1

$KB \models a \vee (b \wedge \sim c) \vee (\sim b \vee c) ?$

Answer:

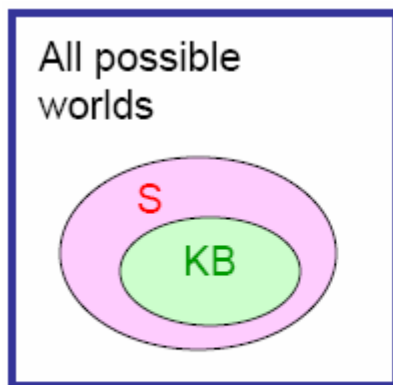
Yes

Because every model that makes  $KB$  True also makes the above statement True.

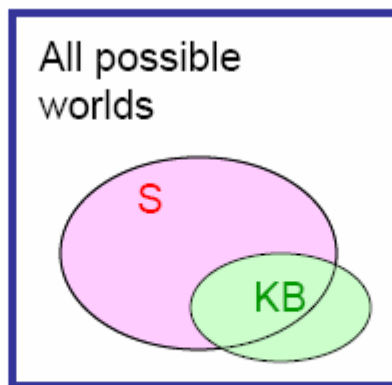


# Logical Entailment

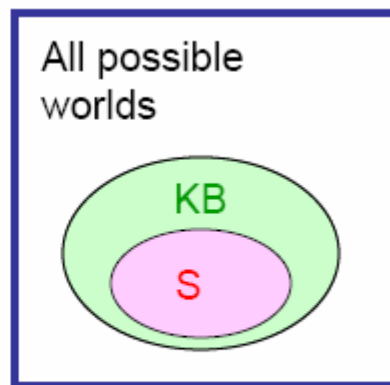
- $KB \models S$  pronounced  
“*knowledge base  $KB$  logically entails  $S$* ”
- And means  
“*all possible models that make  $KB$  evaluate to  $TRUE$  also makes  $S$  evaluate to  $TRUE$* ”



$KB \models S$



~~$KB \models S$~~



~~$KB \models S$~~



# Inference

$KB \vdash_i \alpha$  = sentence  $\alpha$  can be derived from  $KB$  by procedure  $i$

Consequences of  $KB$  are a haystack;  $\alpha$  is a needle.

Entailment = needle in haystack; inference = finding it

Soundness:  $i$  is sound if

whenever  $KB \vdash_i \alpha$ , it is also true that  $KB \models \alpha$

Completeness:  $i$  is complete if

whenever  $KB \models \alpha$ , it is also true that  $KB \vdash_i \alpha$



# Inference by Enumeration

- Enumerate all possible models
- For each model for which KB is true, check that  $\alpha$  is true too.
- Depth-First enumeration  
*Sound and Complete.*
- What is the problem?
- Will grow exponentially with the number of symbols!!
- $O(2^n)$  for  $n$  symbols. Problem is *co-NP-complete*.

$B_{1,1}$	$B_{2,1}$	$P_{1,1}$	$P_{1,2}$	$P_{2,1}$	$P_{2,2}$	$P_{3,1}$	$R_1$	$R_2$	$R_3$	$R_4$	$R_5$	KB
false	false	false	false	false	false	false	true	true	true	true	false	false
false	false	false	false	false	false	true	true	true	false	true	false	false
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
false	true	false	false	false	false	false	true	true	false	true	true	false
false	true	false	false	false	false	true	true	true	true	true	true	true
false	true	false	false	false	true	false	true	true	true	true	true	true
false	true	false	false	false	true	true	true	true	true	true	true	true
false	true	false	false	true	false	false	true	false	false	true	true	false
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
true	true	true	true	true	true	true	false	true	true	false	true	false



## Or... Apply Inference Rules

- Apply inference rules to derive new statements from given ones
- Until... You hit  $\alpha$ .

Also: *Modus-ponens*

$$\begin{aligned}(\alpha \wedge \beta) &\equiv (\beta \wedge \alpha) && \text{commutativity of } \wedge \\(\alpha \vee \beta) &\equiv (\beta \vee \alpha) && \text{commutativity of } \vee \\((\alpha \wedge \beta) \wedge \gamma) &\equiv (\alpha \wedge (\beta \wedge \gamma)) && \text{associativity of } \wedge \\((\alpha \vee \beta) \vee \gamma) &\equiv (\alpha \vee (\beta \vee \gamma)) && \text{associativity of } \vee \\\neg(\neg\alpha) &\equiv \alpha && \text{double-negation elimination} \\(\alpha \Rightarrow \beta) &\equiv (\neg\beta \Rightarrow \neg\alpha) && \text{contraposition} \\(\alpha \Rightarrow \beta) &\equiv (\neg\alpha \vee \beta) && \text{implication elimination} \\(\alpha \Leftrightarrow \beta) &\equiv ((\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)) && \text{biconditional elimination} \\\neg(\alpha \wedge \beta) &\equiv (\neg\alpha \vee \neg\beta) && \text{De Morgan} \\\neg(\alpha \vee \beta) &\equiv (\neg\alpha \wedge \neg\beta) && \text{De Morgan} \\(\alpha \wedge (\beta \vee \gamma)) &\equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) && \text{distributivity of } \wedge \text{ over } \vee \\(\alpha \vee (\beta \wedge \gamma)) &\equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) && \text{distributivity of } \vee \text{ over } \wedge\end{aligned}$$



# Validity and Satisfiability

A sentence is **valid** if it is true in **all** models,

e.g., *True*,  $A \vee \neg A$ ,  $A \Rightarrow A$ ,  $(A \wedge (A \Rightarrow B)) \Rightarrow B$

Validity is connected to inference via the **Deduction Theorem**:

$KB \models \alpha$  if and only if  $(KB \Rightarrow \alpha)$  is valid

A sentence is **satisfiable** if it is true in **some** model

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

A sentence is **unsatisfiable** if it is true in **no** models

e.g.,  $A \wedge \neg A$

Satisfiability is connected to inference via the **Deduction Theorem**:

$KB \models \alpha$  if and only if  $(KB \wedge \neg \alpha)$  is unsatisfiable

i.e., prove  $\alpha$  by *reductio ad absurdum*

Good old “*proof by contradiction*”





# Proof methods

- Proof methods divide into (roughly) two kinds:
- Application of inference rules
  - Legitimate (sound) generation of new sentences from old
  - Proof = a sequence of inference rule applications
  - Can use inference rules as operators in a standard search alg.
  - Typically require translation of sentences into a normal form
- Model checking
  - truth table enumeration (always exponential in  $n$ )
  - improved backtracking, e.g., Davis-Putnam-Logemann-Loveland
  - heuristic search in model space (sound but incomplete)
  - e.g., min-conlicts-like hill-climbing algorithms



# Forward and Backward Chaining

Horn Form (restricted)

KB = **conjunction** of **Horn clauses**

Horn clause =

◇ proposition symbol; or

◇ (conjunction of symbols)  $\Rightarrow$  symbol

E.g.,  $C \wedge (B \Rightarrow A) \wedge (C \wedge D \Rightarrow B)$

Modus Ponens (for Horn Form): complete for Horn KBs

$$\frac{\alpha_1, \dots, \alpha_n, \quad \alpha_1 \wedge \dots \wedge \alpha_n \Rightarrow \beta}{\beta}$$

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



# Example: Forward Chaining

Idea: fire any rule whose premises are satisfied in the *KB*,  
add its conclusion to the *KB*, until query is found

$$P \Rightarrow Q$$

$$L \wedge M \Rightarrow P$$

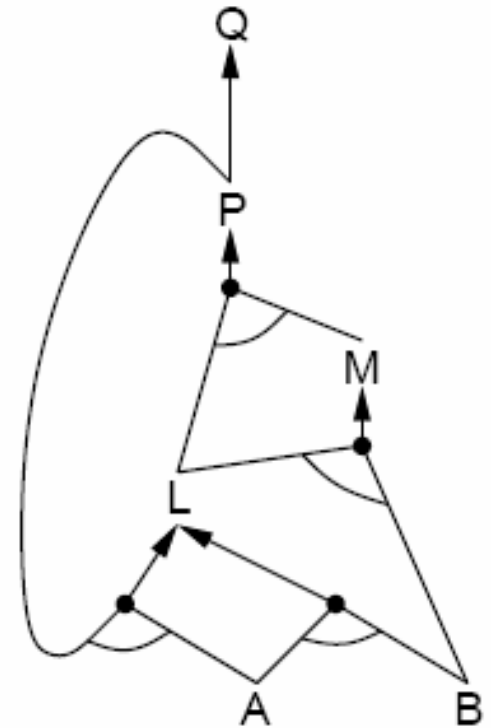
$$B \wedge L \Rightarrow M$$

$$A \wedge P \Rightarrow L$$

$$A \wedge B \Rightarrow L$$

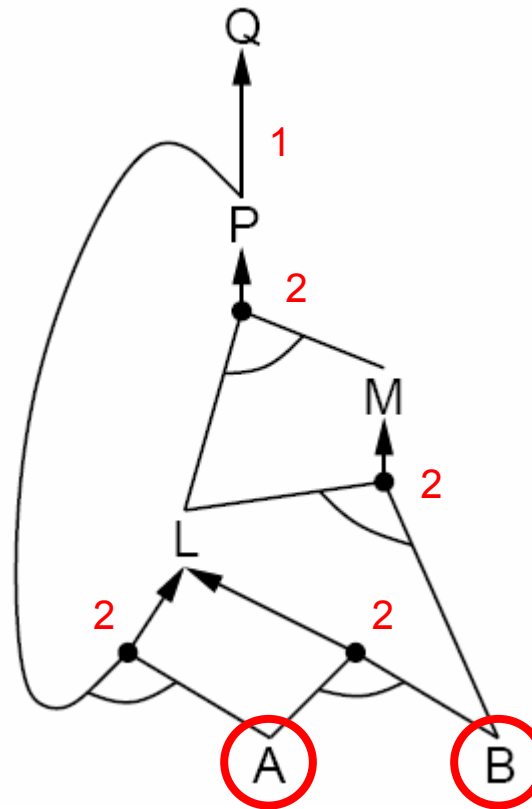
*A*

*B*





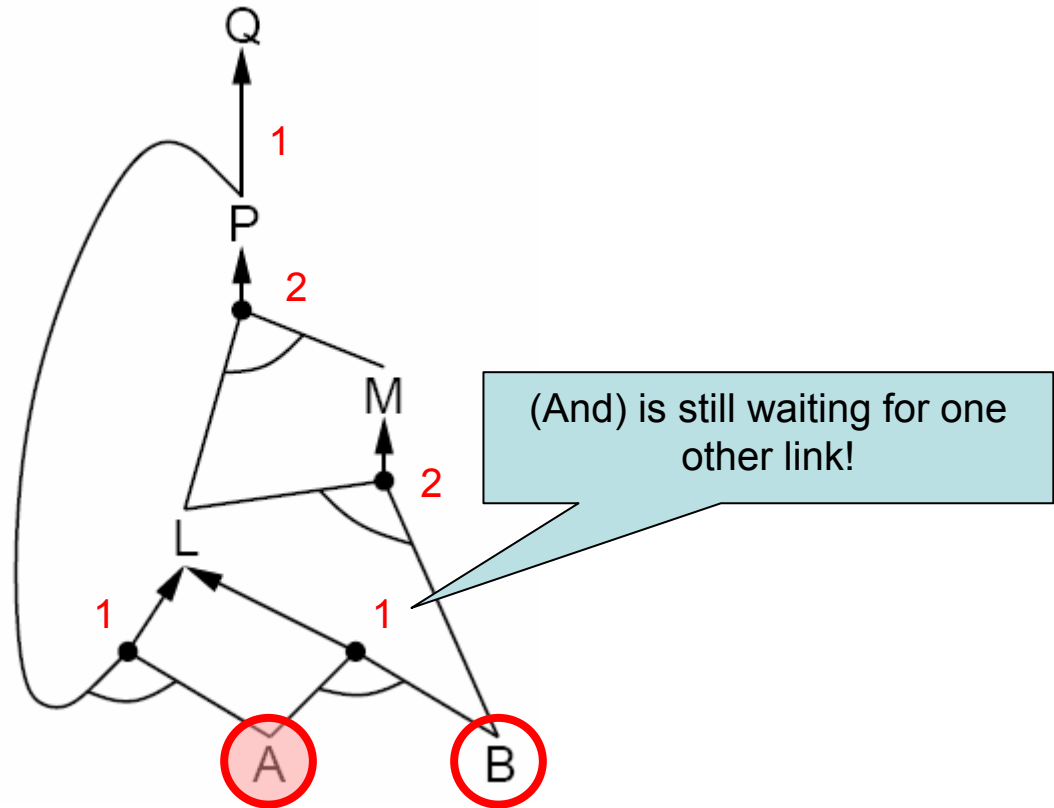
# Example: Forward Chaining



We start with facts in the KB.

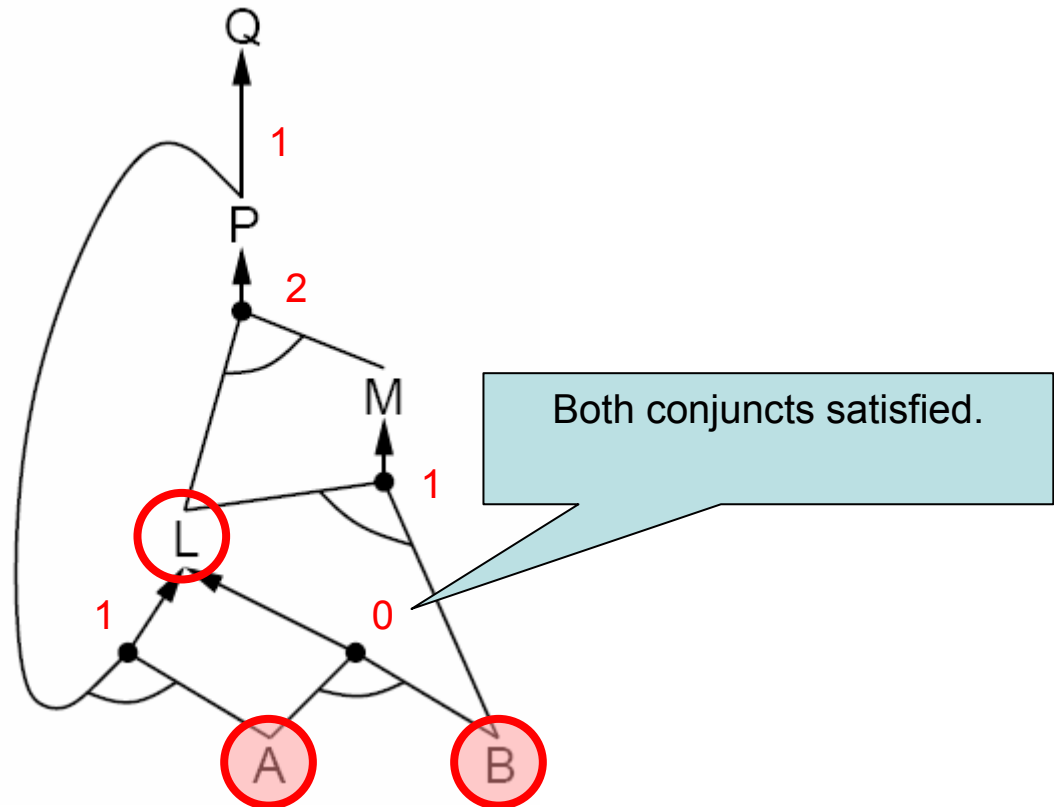


# Example: Forward Chaining



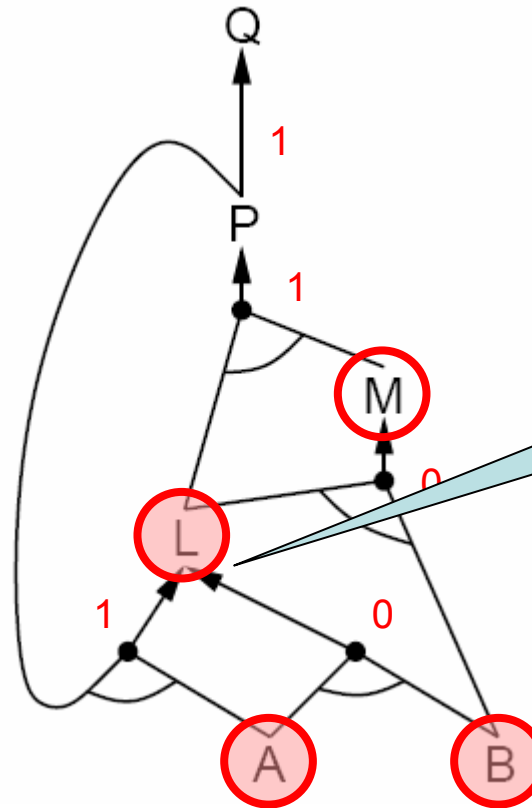


# Example: Forward Chaining



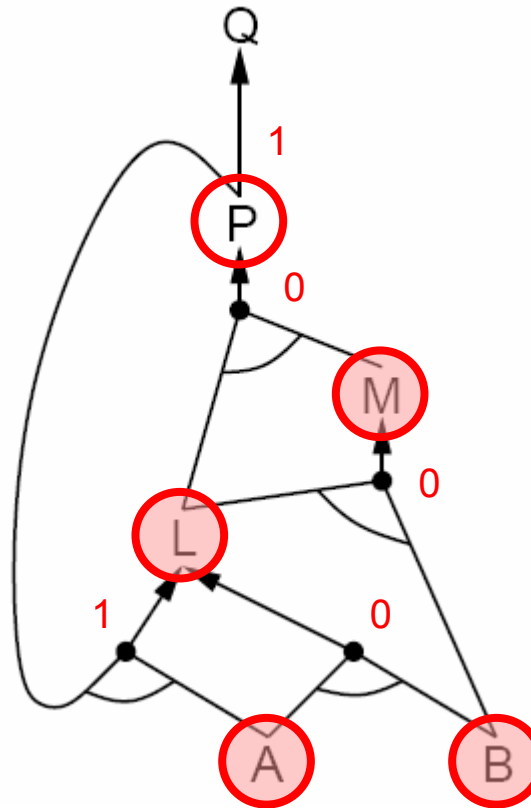


# Example: Forward Chaining





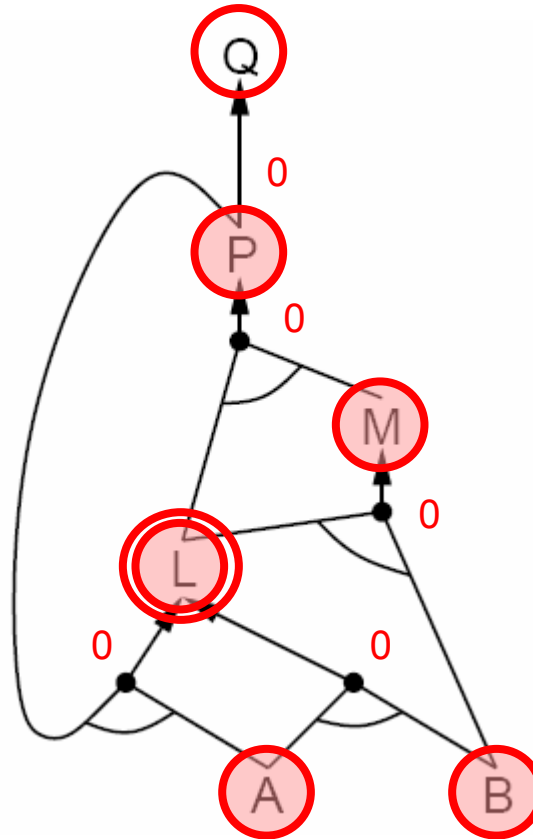
# Example: Forward Chaining





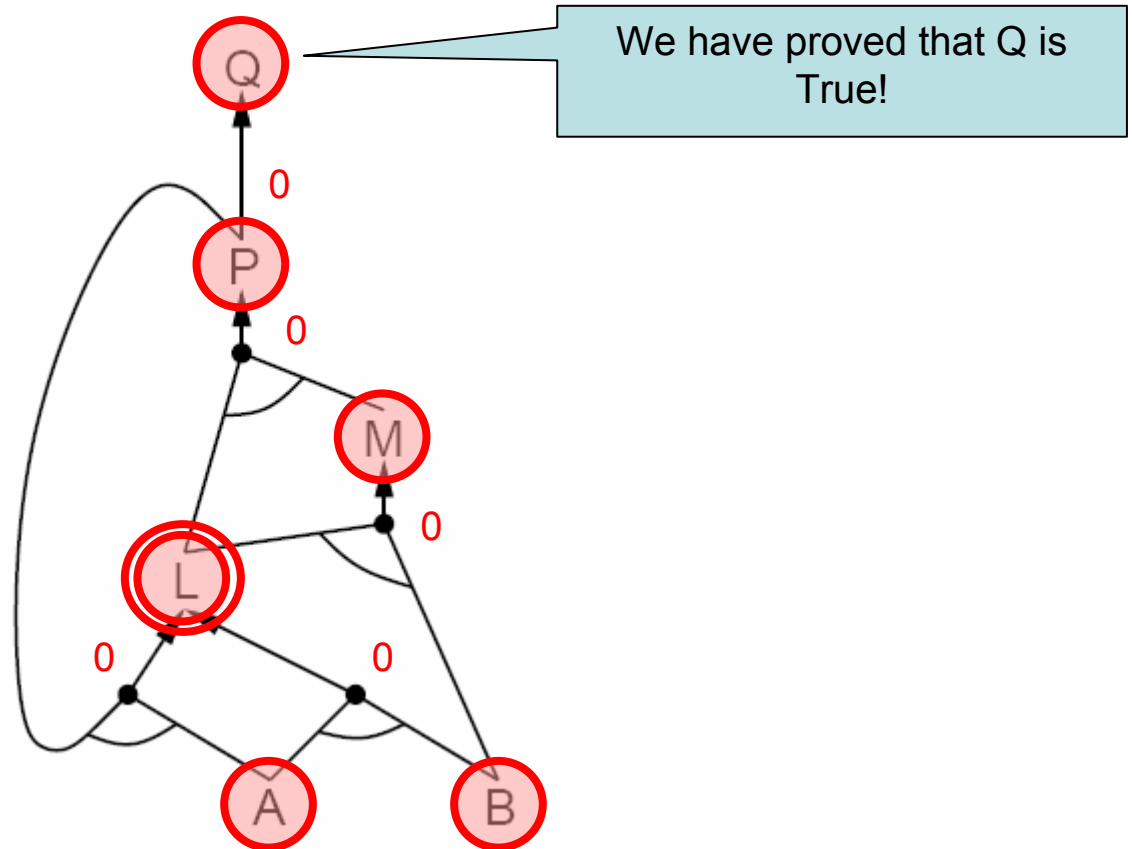


# Example: Forward Chaining





# Example: Forward Chaining



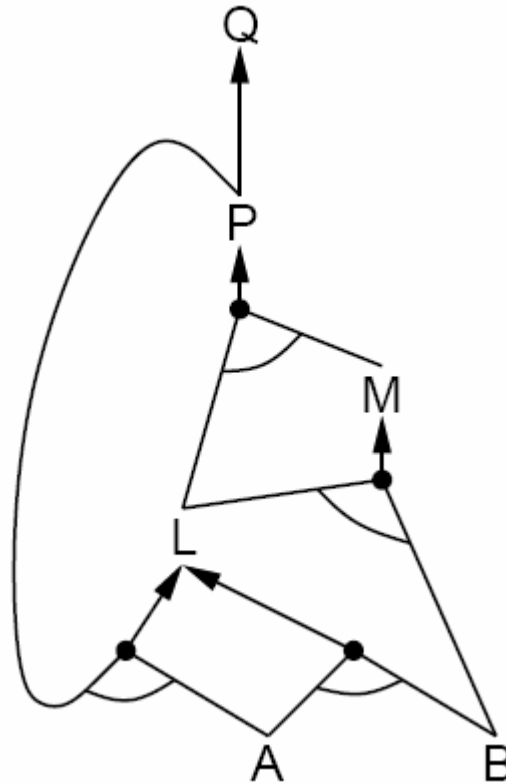


# Backward Chaining

- 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 subgoal is already on the goal stack
- Avoid repeated work: check if new subgoal
  - 1) has already been proved true, or
  - 2) has already failed



# Backward Chaining





# Forward vs Backward Chaining

- FC is **data-driven**, cf. 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