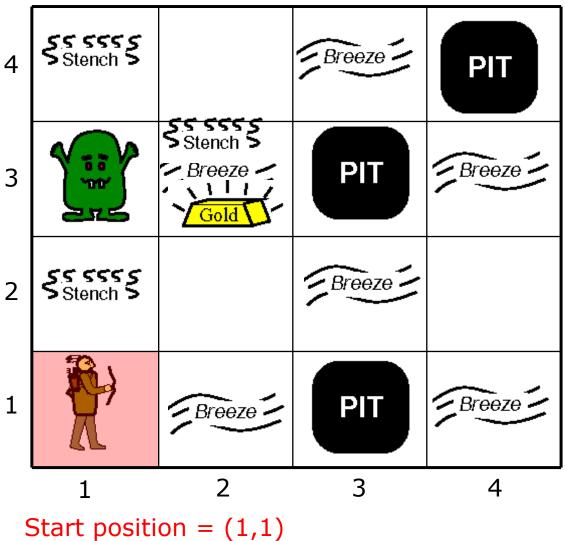
### Cooperating Intelligent Systems

Logical agents
Chapter 7, AIMA

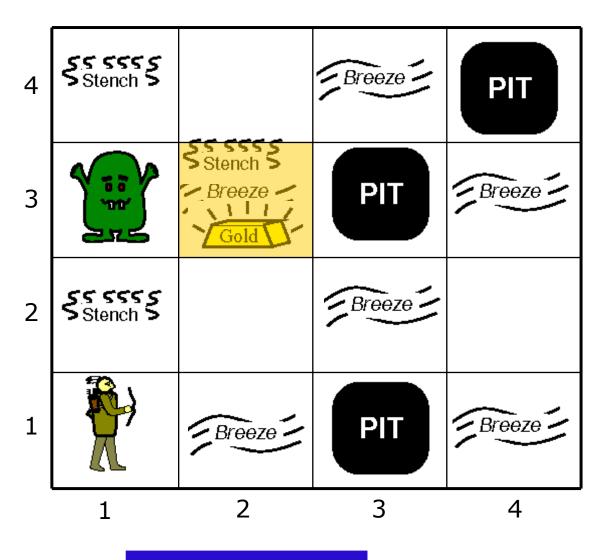
## Motivation for Knowledge Representation

- Search algorithms discussed previously can be called meta-programming
  - but it still is programming
  - the code needs to be specialised for every concrete application according
- We need something more general
  - specify only the rules of the game
  - use "out-of-the-box" reasoning engine

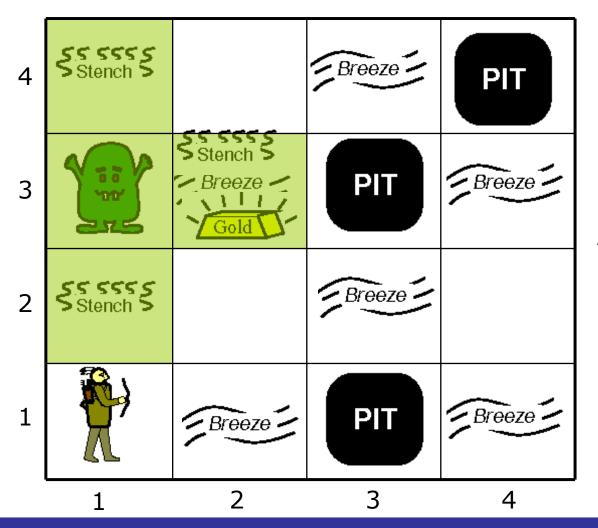
4	SS SSS S Stench		Breeze	PIT
3		Stench S Breeze	Ξ	Breeze
2	SS SSS S Stench		Breeze	
1		Breeze	PIT	Breeze
•	1	2	3	4



Start position = (1,1)Always safe

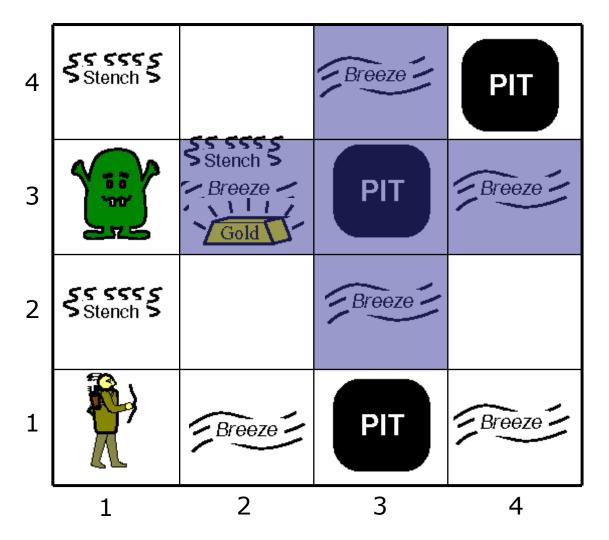


Goal: Get the gold



The environment is static: the Wumpus doesn't move around

Problem 1: Big, hairy, smelly, dangerous Wumpus. Will eat you if you run into it, but you can smell it a block away.



Problem 2: Big, bottomless pits where you fall down. You can feel the breeze when you are near them.

### PEAS description

#### **Performance measure:**

- +1000 for gold
- -1000 for being eaten or falling down pit
- -1 for each action
- -10 for using the arrow

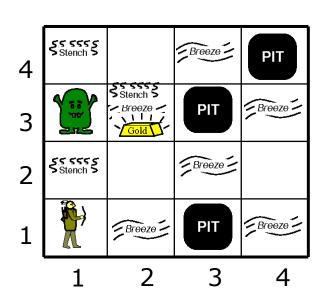
#### **Environment:**

4×4 grid of "rooms", each "room" can be empty, with gold, occupied by Wumpus, or with a pit.

#### **Acuators:**

Move forward, turn left 90°, turn right 90° Grab, shoot

#### **Sensors:**



### PEAS description

#### **Performance measure:**

- +1000 for gold
- -1000 for being eaten or falling down pit
- -1 for each action
- -10 for using the arrow

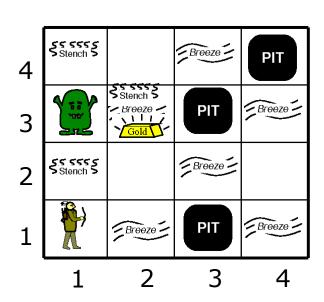
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Move forward, turn left 90°, turn right 90° Grab, shoot

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### PEAS description

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#### **Environment:**

4×4 grid of "rooms", each "room" can be empty, with gold, occupied by Wumpus, or with a pit.

#### **Acuators:**

Move forward, turn left 90°, turn right 90° Grab, shoot

#### **Sensors:**

$$\begin{pmatrix} \alpha_{1} \\ \alpha_{2} \\ \alpha_{3} \\ \alpha_{4} \\ \alpha_{5} \end{pmatrix} = \begin{pmatrix} forward \\ turn \ left \\ turn \ right \\ grab \\ shoot \end{pmatrix}, \alpha_{i} \in \{0,1\}$$

### PEAS description

#### **Performance measure:**

- +1000 for gold
- -1000 for being eaten or falling down pit
- -1 for each action
- -10 for using the arrow

#### **Environment:**

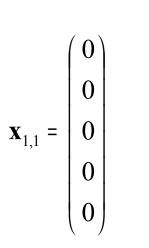
4×4 grid of "rooms", each "room" can be empty, with gold, occupied by Wumpus, or with a pit.

#### **Actuators:**

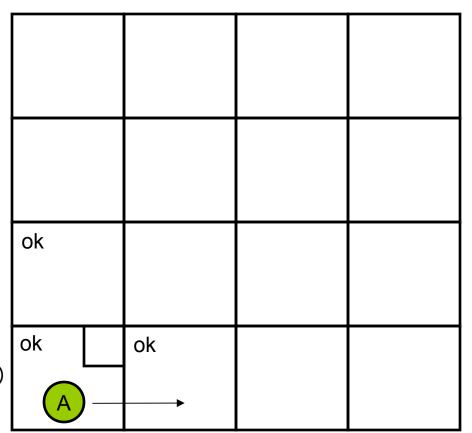
Move forward, turn left 90°, turn right 90° Grab, shoot

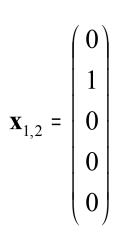
#### **Sensors:**

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \begin{pmatrix} stench \\ breeze \\ wall \\ glitter \\ scream \end{pmatrix}, x_i \in \{0,1\}$$

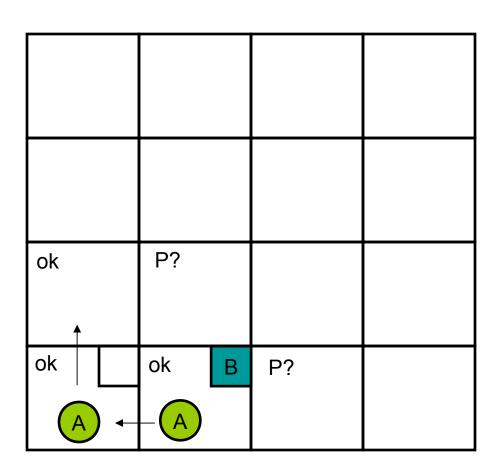


Agent senses nothing (no breeze, no smell,..)



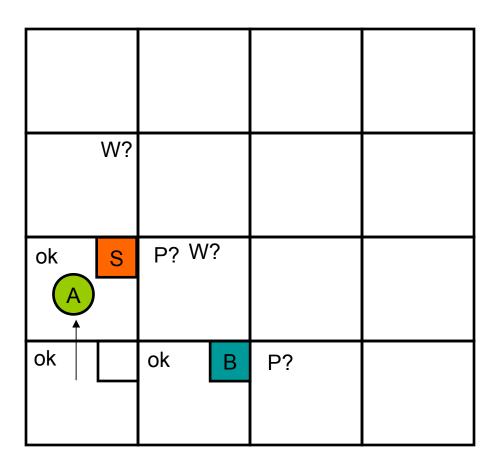


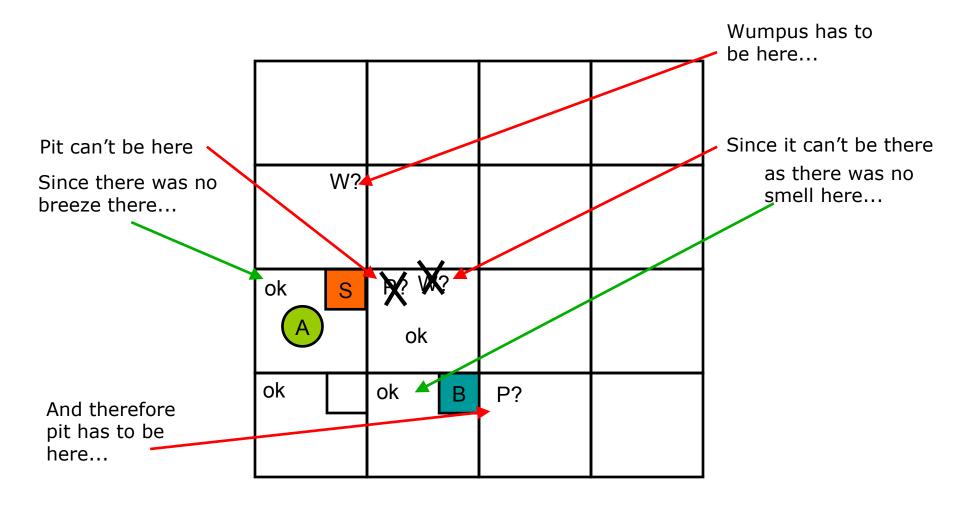
Agent feels a breeze



Agent feels a foul smell

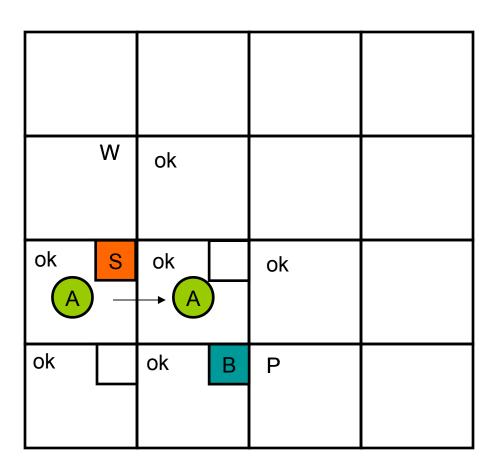
$$\mathbf{x}_{2,1} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$





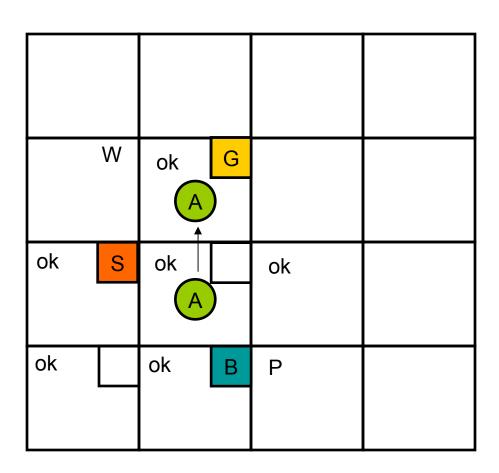
Agent senses nothing (no breeze, no smell,..)

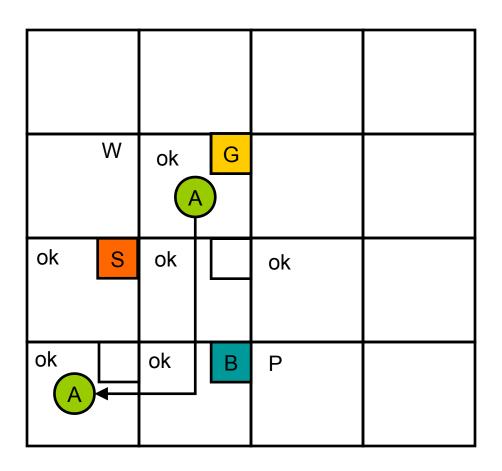
$$\mathbf{x}_{2,2} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$



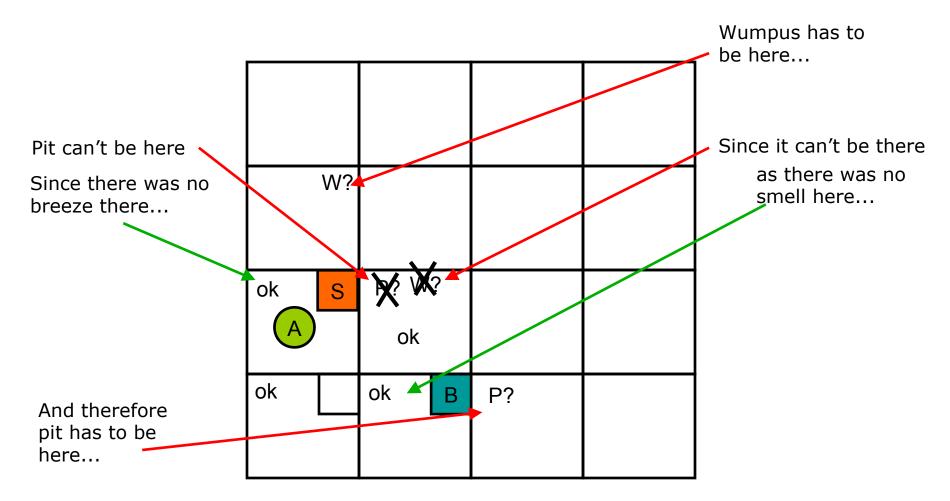
Agent senses breeze, smell, and sees gold!

$$\mathbf{x} = \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 0 \end{pmatrix}$$





Grab the gold and get out!



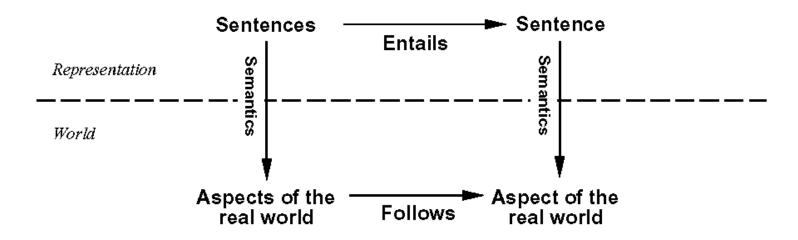
How do we automate this kind of reasoning? How can we make computers figure it out on their own?

## Logic

Logic is a formal language for representing information in such a way that conclusions can be drawn

### A logic has

- Syntax that specifies symbols in the language and how they can be combined to form sentences
- **Semantics** that specifies what facts in the world these sentences refer to and assigns *truth values* to them based on their meaning in the world.
- Inference procedure, a mechanical method for computing (deriving) new (true) sentences from existing (known) sentences.



### Entailment

 $A \models B$ 

The sentence A entails the sentence B

- If A is true, then B must also be true
- B is a "logical consequence" of A

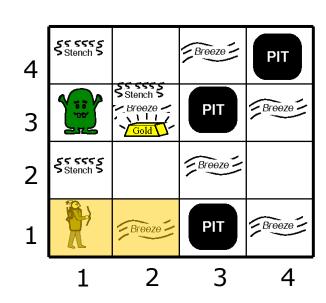
Let's explore this concept a bit...

## Example: Wumpus entailment

Agent's knowledge base (KB) after having visited (1,1) and (1,2):

- The rules of the game (PEAS)
- 2) Nothing in (1,1)
- 3) Breeze in (1,2)

Which models (states of the world) match these observations?



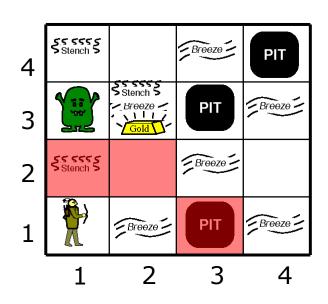
$$\mathbf{x}_{1,1} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \quad \mathbf{x}_{1,2} = \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

## Example: Wumpus entailment

We only care about neighboring rooms, i.e. {(2,1),(2,2),(1,3)}. We can't know anything about the other rooms.

We care about pits, because we have detected a breeze. We don't want to fall down a pit.

There are 2<sup>3</sup>=8 possible arrangements of {pit, no pit} in the three neighboring rooms.

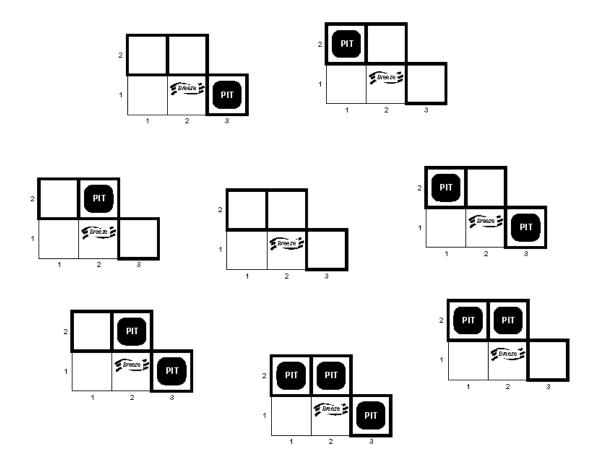


### Possible conclusions:

 $\alpha_{l}$ : There is no pit in (2,1)

 $\alpha_2$ : There is no pit in (2,2)

 $\alpha_3$ : There is no pit in (1,3)

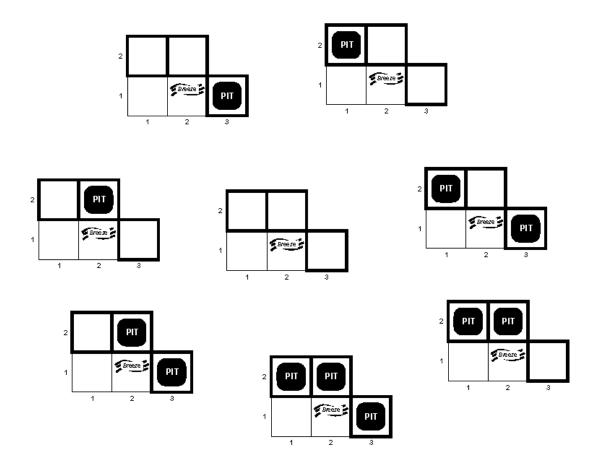


The eight possible situations...



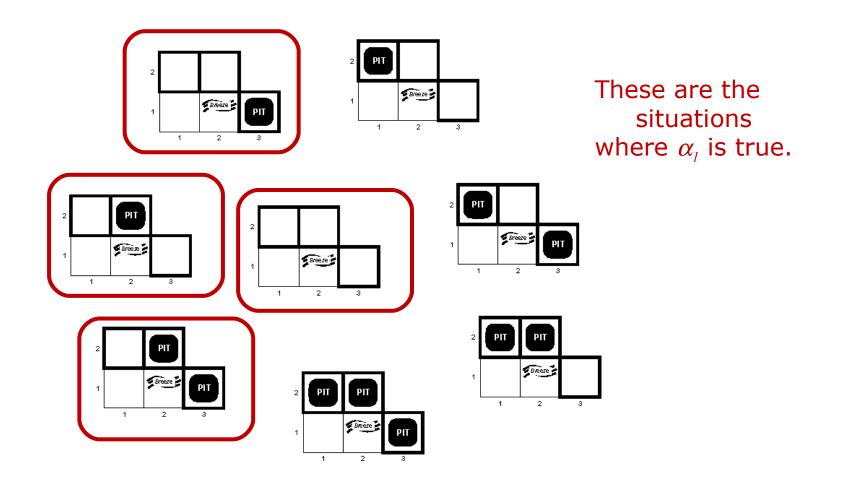
The eight possible situations...

These are the ones that agree with our Knowledge Base (KB), i.e. the rules of the game and our observations.



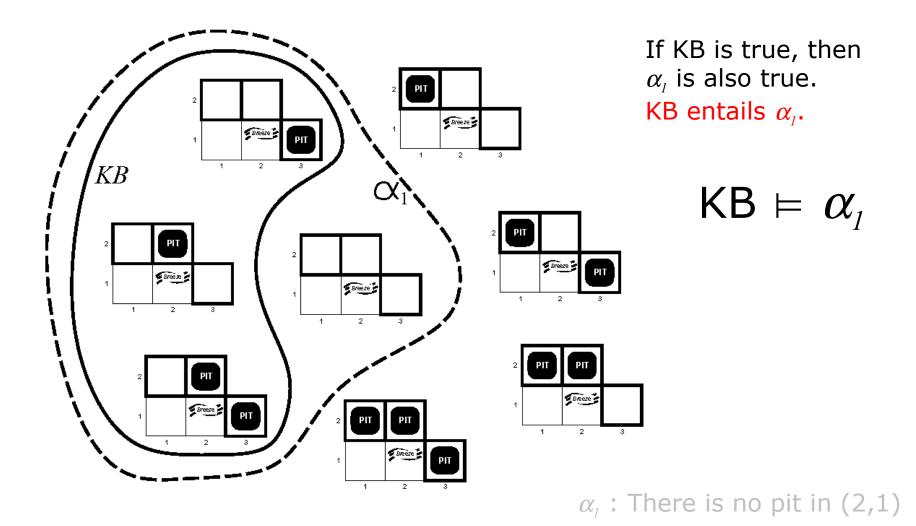
 $\alpha_{i}$ : There is no pit in (2,1)

...let's explore this conclusion



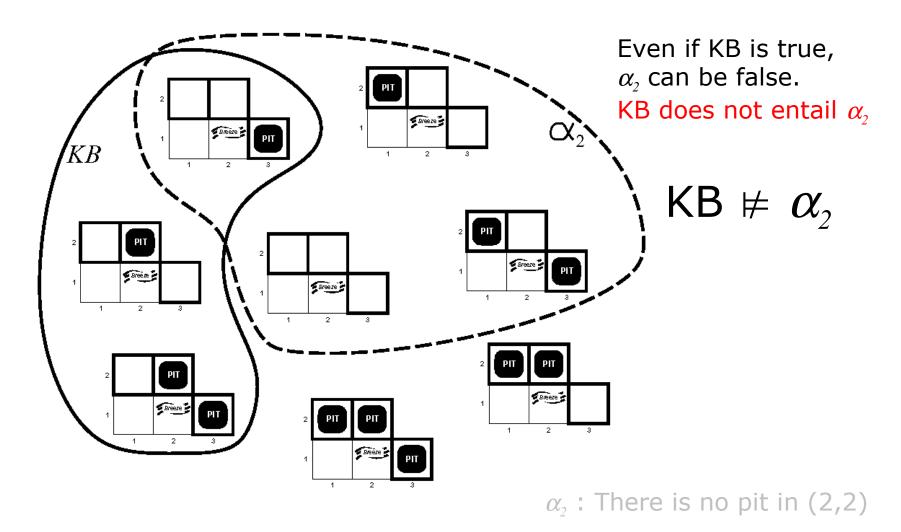
 $\alpha_i$ : There is no pit in (2,1)

...let's explore this conclusion



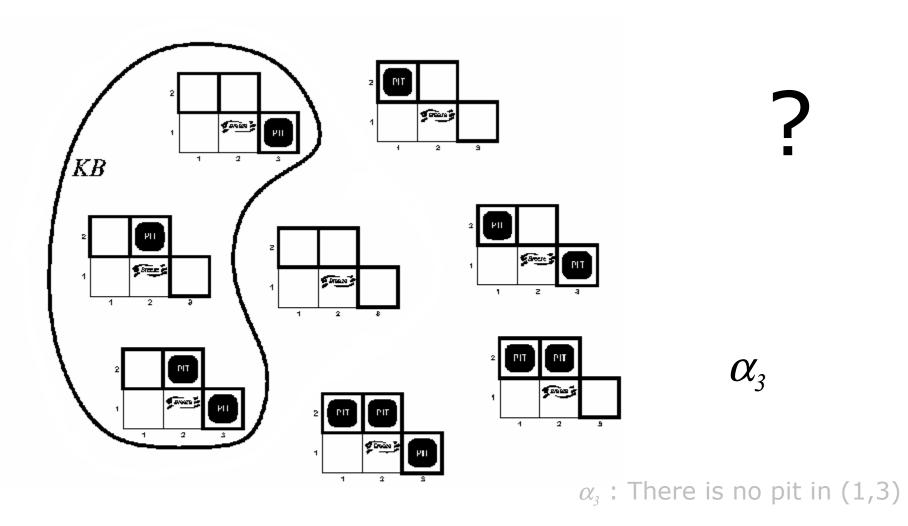
KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 $\alpha_I$  = The set of models that agrees with conclusion  $\alpha_I$  [conclusion  $\alpha_I$  is true in these models]



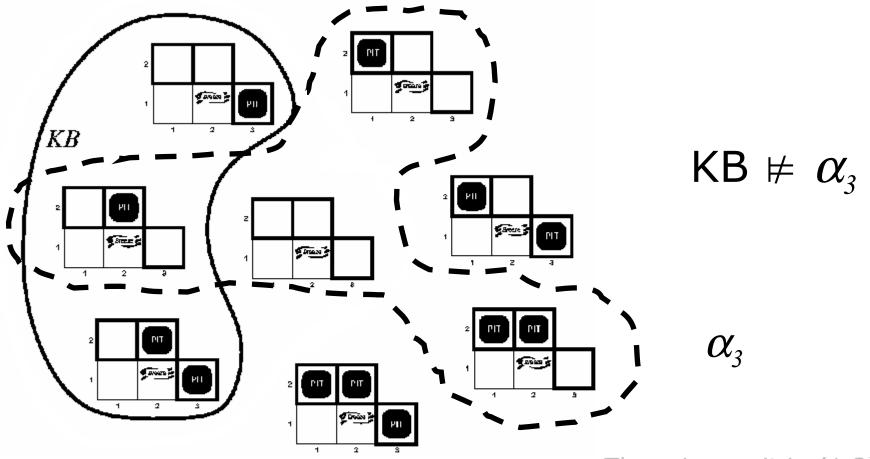
KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 $\alpha_2$  = The set of models that agrees with conclusion  $\alpha_2$  [conclusion  $\alpha_2$  is true in these models]



KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 $\alpha_3$  = The set of models that agrees with conclusion  $\alpha_3$  [conclusion  $\alpha_3$  is true in these models]



 $\alpha_3$ : There is no pit in (1,3)

KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 $\alpha_3$  = The set of models that agrees with conclusion  $\alpha_3$  [conclusion  $\alpha_3$  is true in these models]

## Inference engine

- We need an algorithm that produces the entailed conclusions <u>automatically</u>
  - for any user-defined Knowledge Base
- Entailment is the most important property in logic
  - most interesting things can be expressed using entailment
- We will call this algorithm, and it's implementation, an inference engine

## Inference engine

Inference engine domain-independent algorithms

Knowledge base domain-specific content

$$KB \vdash_{i} A$$

"A is derived from KB by inference engine i"

- **Truth-preserving:** *i* only derives entailed sentences
- Complete: i derives all entailed sentences

We want inference engines that are both truth-preserving and complete

## Propositional (boolean) logic Syntax

**Atomic sentence** = a single propositional symbol e.g. P, Q,  $P_{13}$ ,  $W_{31}$ ,  $G_{32}$ , T, F

**Complex sentence** = combination of simple sentences using *connectives* 

- ¬ (not) negation
- ∧ (and) conjunction
- v (or) disjunction
- ⇒ (implies) implication

$$P_{13}\,\wedge\,W_{31}$$

$$W_{31} \vee \neg W_{31}$$

$$W_{31} \Rightarrow S_{32}$$

⇔ (iff = if and only if) biconditional/logical equality

Precedence:  $\neg, \land, \lor, \Rightarrow, \Leftrightarrow$ 

## Propositional (boolean) logic Semantics

**Semantics:** The rules for whether a sentence is true or false

- T (true) is true in every model
- F (false) is false in every model
- The truth values for other proposition symbols are specified in the model

Atomic sentences

- Truth values for complex sentences are specified according to the definitions of connectives
  - using a truth table

### Boolean truth table

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔Q
False	False					
False	True					
True	False					
True	True					

Please complete this table...

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔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

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔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

Not P is the opposite of P

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔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

 $P \wedge Q$  is true only when both P and Q are true

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔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

 $P \lor Q$  is true when either P or Q is true

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔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

 $P \Rightarrow Q$ : If P is true then we claim that Q is true, otherwise we make no claim

Р	Q	¬P	P∧Q	P∨Q	P⇒Q	P⇔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

 $P \Leftrightarrow Q$  is true when the truth values for P and Q are identical

Р	Q	P⊕Q
False	False	
False	True	
True	False	
True	True	

The exlusive or (XOR) is different from the OR

Р	Q	P⊕Q
False	False	False
False	True	True
True	False	True
True	True	False

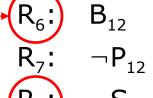
The exlusive or (XOR) is different from the OR

# Example: Wumpus KB

Interesting sentences [tell us what is in <a href="neighbour">neighbour</a> squares]

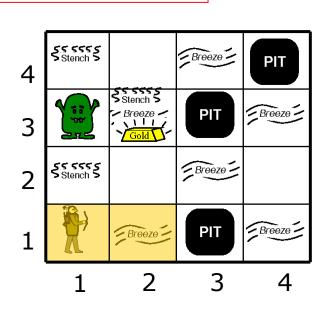
#### Knowledge base

$$R_{1}$$
:  $\neg P_{11}$ 
 $R_{2}$ :  $\neg B_{11}$ 
 $R_{3}$ :  $\neg W_{11}$ 
 $R_{4}$ :  $\neg S_{11}$ 
 $R_{5}$ :  $\neg G_{11}$ 



$$R_9$$
:  $\neg W_{12}$ 

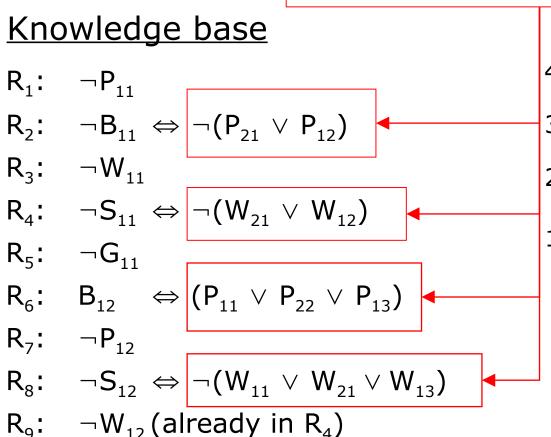
$$R_{10}$$
:  $\neg G_{12}$ 

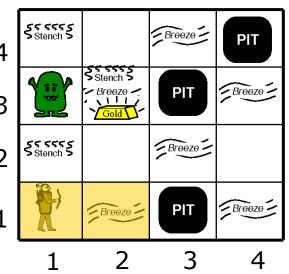


- 1. Nothing in (1,1)
- 2. Breeze in (1,2)

# Example: Wumpus KB

We infer this from the rules of the game





- 1. Nothing in (1,1)
- 2. Breeze in (1,2)

$$R_{10}$$
:  $\neg G_{12}$ 

$$\mathsf{KB} = \mathsf{R}_1 \, \wedge \, \mathsf{R}_2 \, \wedge \, \mathsf{R}_3 \, \wedge \, \mathsf{R}_4 \, \wedge \, \mathsf{R}_5 \, \wedge \, \mathsf{R}_6 \, \wedge \, \mathsf{R}_7 \, \wedge \, \mathsf{R}_8 \, \wedge \, \mathsf{R}_9 \, \wedge \, \mathsf{R}_{10}$$

What is in squares (1,3), (2,1), and (2,2)?

#	W <sub>21</sub>	W <sub>22</sub>	W <sub>13</sub>	P <sub>21</sub>	P <sub>22</sub>	P <sub>13</sub>	$R_2$	$R_4$	$R_6$	R <sub>8</sub>	
1	0	0	0	0	0	0	1	1	0	1	
2	0	0	0	0	0	1	1	1	1	1	
3	0	0	0	0	1	0	1	1	1	1	KB true
4	0	0	0	0	1	1	1	1	1	1	
5	0	0	0	1	0	0	0	1	0	1	
:	:	:	:	i	:	:	:	:	:	1	
63	0	1	1	1	1	1	0	1	1	0	
64	1	1	1	1	1	1	0	0	1	0	
									K	B	

(interesting sentences)

We have 6 interesting sentences:  $W_{21}$ ,  $W_{22}$ ,  $W_{13}$ ,  $P_{21}$ ,  $P_{22}$ ,  $P_{13}$ :  $2^6 \neq 64$  comb.

What is in squares (1,3), (2,1), and (2,2)?

#	$W_{21}$	W <sub>22</sub>	W <sub>13</sub>	P <sub>21</sub>	P <sub>22</sub>	P <sub>13</sub>	$R_2$	$R_4$	$R_6$	R <sub>8</sub>		
1	0	0	0	0	0	0	1	1	0	1	_	
2	0	0	0	0	0	1	1	1	1	1		
3	0	0	0	0	1	0	1	1	1	1	<b> </b>	KB true
4	0	0	0	0	1	1	1	1	1	1		
5	0	0	0	1	0	0	0	1	0	1		
:	:	1	:	:	:	:	1	1	:	:		
63	0	1	1	1	1	1	0	1	1	0		
64	1	1	1	1	1	1	0	0	1	0		

What do we deduce from this?

What is in squares (1,3), (2,1), and (2,2)?

#	$W_{21}$	W <sub>22</sub>	W <sub>13</sub>	P <sub>21</sub>	P <sub>22</sub>	P <sub>13</sub>	$R_2$	$R_4$	$R_6$	R <sub>8</sub>	
1	0	0	0	0	0	0	1	1	0	1	_
2	0	0	0	0	0	1	1	1	1	1	
3	0	0	0	0	1	0	1	1	1	1	├ KB ⊤true
4	0	0	0	0	1	1	1	1	1	1	
5	0	0	0	1	0	0	0	1	0	1	
:	:	:	:	:	:	1	:	:	:	:	
63	0	1	1	1	1	1	0	1	1	0	
64	1	1	1	1	1	1	0	0	1	0	

$$KB \models \neg W_{21} \land \neg W_{22} \land \neg W_{13} \land \neg P_{21}$$

- Can be implemented as a depth-first search on a constraint graph
  - with backtracking
- Time complexity  $\sim O(2^n)$  where n is the number of relevant symbols
- Space complexity  $\sim O(n)$

Not very efficient...

Although computers are good with long sequences of 0s and 1s

### Some more definitions

#### **Equivalence:**

$$A \equiv B \text{ iff } A \models B \text{ and } B \models A$$

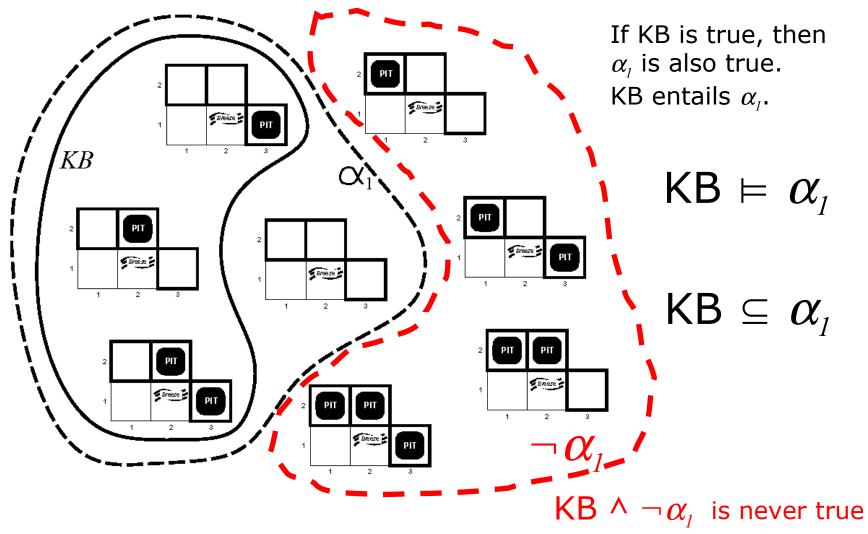
Validity: A valid sentence is true in all models (a tautology)

$$A \models B \text{ iff } (A \Rightarrow B) \text{ is valid}$$

Satisfiability: A sentence is satisfiable if it is true in some model

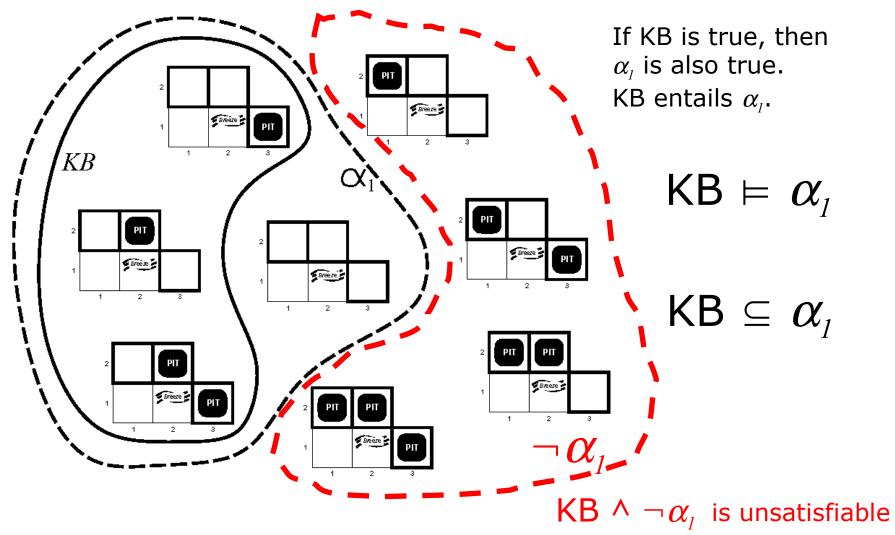
$$A \models B \text{ iff } (A \land \neg B) \text{ is unsatisfiable}$$

Let's explore *satisfiability* first...



KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 $\alpha_I$  = The set of models that agrees with conclusion  $\alpha_I$  [conclusion  $\alpha_I$  is true in these models]



KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 $\alpha_I$  = The set of models that agrees with conclusion  $\alpha_I$  [conclusion  $\alpha_I$  is true in these models]

### Some more definitions

#### **Equivalence:**

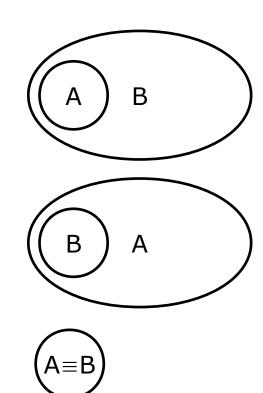
 $A \equiv B \text{ iff } A \models B \text{ and } B \models A$ 

For example,  $A \equiv \neg(\neg A)$ 

 $A \models B$  means that the set of models where A is true is a subset of the models where B is true:  $A \subseteq B$ 

 $B \models A$  means that the set of models where B is true is a subset of the models where A is true:  $B \subseteq A$ 

Therefore, the set of models where A is true must be equal to the set of models where B is true:  $A \equiv B$ 

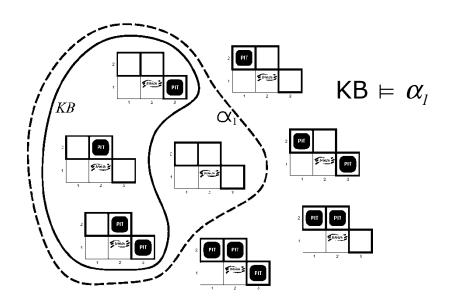


### Some more definitions

**Validity:** A valid sentence is true in all models (a tautology)

For example,  $A \vee \neg A$  is valid

 $A \models B \text{ iff } (A \Rightarrow B) \text{ is valid}$ 



А	В	A⇒B
False	False	True
False	True	True
True	False	False
True	True	True

## Logical equivalences

$$(A \wedge B) \equiv (B \wedge A) \qquad \wedge \text{ is commutative} \\ (A \vee B) \equiv (B \vee A) \qquad \vee \text{ is commutative} \\ ((A \wedge B) \wedge C) \equiv (A \wedge (B \wedge C)) \qquad \wedge \text{ is associative} \\ ((A \vee B) \vee C) \equiv (A \vee (B \vee C)) \qquad \vee \text{ is associative} \\ \neg (\neg A) \equiv A \qquad \qquad \text{Double-negation elimination} \\ (A \Rightarrow B) \equiv (\neg B \Rightarrow \neg A) \qquad \text{Contraposition} \\ (A \Rightarrow B) \equiv (\neg A \vee B) \qquad \text{Implication elimination} \\ (A \Leftrightarrow B) \equiv ((A \Rightarrow B) \wedge (B \Rightarrow A)) \qquad \text{Biconditional elimination} \\ \neg (A \wedge B) \equiv (\neg A \vee \neg B) \qquad \text{"De Morgan"} \\ \neg (A \vee B) \equiv (\neg A \wedge \neg B) \qquad \text{"De Morgan"} \\ (A \wedge (B \vee C)) \equiv ((A \wedge B) \vee (A \wedge C)) \qquad \text{Distributivity of } \wedge \text{ over } \vee \\ (A \vee (B \wedge C)) \equiv ((A \vee B) \wedge (A \vee C)) \qquad \text{Distributivity of } \vee \text{ over } \wedge \\ \end{pmatrix}$$

А	В	A ∧ B	¬(A ∧ B)	$\neg A$	¬В	$\neg A \lor \neg B$
False	False					
False	True					
True	False					
True	True					

А	В	A ∧ B	¬(A ∧ B)	$\neg A$	¬В	$\neg A \lor \neg B$
False	False	False				
False	True	False				
True	False	False				
True	True	True				

А	В	A ∧ B	¬(A ∧ B)	$\neg A$	¬В	$\neg A \lor \neg B$
False	False	False	True			
False	True	False	True			
True	False	False	True			
True	True	True	False			

А	В	A ∧ B	¬(A ∧ B)	$\neg A$	¬В	$\neg A \lor \neg B$
False	False	False	True	True	True	
False	True	False	True	True	False	
True	False	False	True	False	True	
True	True	True	False	False	False	

А	В	A ∧ B	¬(A ∧ B)	¬A	¬В	$\neg A \lor \neg B$
False	False	False	True	True	True	True
False	True	False	True	True	False	True
True	False	False	True	False	True	True
True	True	True	False	False	False	False

## Logical equivalences

Work out some of these on paper for yourself, before we move on...

#### Inference

- There are two main approaches towards inference:
  - model enumeration
  - inference rules

#### Inference rules

Inference rules are written as

Antecedent Consequent

If the KB contains the antecedent, you can add the consequent (the KB entails the consequent)

#### Inference rules

Inference rules are written as

If the KB contains the antecedent, you can add the consequent (the KB entails the consequent)

### Commonly used inference rules

#### **Modus Ponens**

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

**Modus Tolens** 

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

**Unit Resolution** 

$$\frac{A \lor B, \neg B}{A}$$

And Elimination

$$\frac{A \wedge B}{A}$$

Or introduction

$$\frac{A}{A \lor B}$$

And introduction

$$\frac{A,B}{A \wedge B}$$

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

	А	В	$A \Rightarrow B$
1	False	False	True
2	False	True	True
3	True	False	False
4	True	True	True

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

	А	В	$A \Rightarrow B$
1	False	False	True
2	False	True	True
3	True	False	False
4	True	True	True

These are the cases when A is True

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

	А	В	$A \Rightarrow B$
1	False	False	True
2	False	True	True
3	True	False	False
4	True	True	True



These are the cases when A ⇒ B is True

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

	Α	В	$A \Rightarrow B$
1	False	False	True
2	False	True	True
3	True	False	False
4	True	True	True

This is the case when both A and A ⇒ B is True

B is also True here so we can safely add B = True to our KB

### Proof for Unit Resolution

$$\frac{A \vee B, \neg B}{A}$$

	Α	В	A ∨ B	$\neg A$	¬В
1	False	False	False	True	True
2	False	True	True	True	False
3	True	False	True	False	True
4	True	True	True	False	False

### Proof for Unit Resolution

$$\frac{A \vee B, \neg B}{A}$$

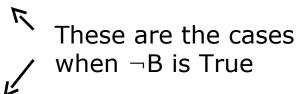
	Α	В	A ∨ B	$\neg A$	¬В
1	False	False	False	True	True
2	False	True	True	True	False
3	True	False	True	False	True
4	True	True	True	False	False

These are the cases when A  $\vee$  B is True

### Proof for Unit Resolution

$$\frac{A \vee B, \neg B}{A}$$

	Α	В	A ∨ B	$\neg A$	¬В
1	False	False	False	True	True
2	False	True	True	True	False
3	True	False	True	False	True
4	True	True	True	False	False



### Proof for Unit Resolution

$$\frac{A \lor B, \neg B}{A}$$

	Α	В	A ∨ B	$\neg A$	¬В
1	False	False	False	True	True
2	False	True	True	True	False
3	True	False	True	False	True
4	True	True	True	False	False

This is the case when both  $\neg B$  and  $A \lor B$  are True

A is also True here so we can safely add A = True to our KB

### Commonly used inference rules

#### **Modus Ponens**

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

Modus Tolens

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

Unit Resolution

$$\frac{A \lor B, \neg B}{A}$$

And Elimination

$$\frac{A \wedge B}{A}$$

Or introduction

$$\frac{A}{A \lor B}$$

And introduction

$$\frac{A,B}{A \wedge B}$$

### Example: Proof in Wumpus KB

#### Knowledge base

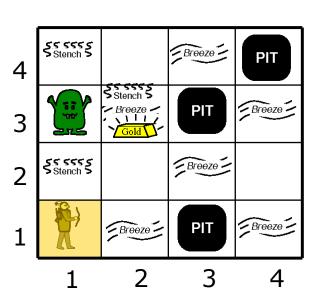
```
R_1: \neg P_{11}
```

$$R_2$$
:  $\neg B_{11}$ 

 $R_3$ :  $\neg W_{11}$ 

 $R_4$ :  $\neg S_{11}$ 

 $R_5$ :  $\neg G_{11}$ 



1. Nothing in (1,1)

$B_{11} \Leftrightarrow (P_{12} \vee P_{21})$	Rule of the game
$B_{11} \Rightarrow (P_{12} \vee P_{21}) \wedge (P_{12} \vee P_{21}) \Rightarrow B_{11}$	Biconditional elimination

$$(A \Leftrightarrow B) \equiv ((A \Rightarrow B) \land (B \Rightarrow A))$$

$$\frac{\mathsf{B}_{11} \Rightarrow (\mathsf{P}_{12} \vee \mathsf{P}_{21}) \wedge (\mathsf{P}_{12} \vee \mathsf{P}_{21}) \Rightarrow \mathsf{B}_{11}}{(\mathsf{P}_{12} \vee \mathsf{P}_{21}) \Rightarrow \mathsf{B}_{11}} \qquad \begin{array}{c} \mathsf{Biconditional} \\ \mathsf{elimination} \end{array}$$
$$(\mathsf{P}_{12} \vee \mathsf{P}_{21}) \Rightarrow \mathsf{B}_{11} \qquad \qquad \mathsf{And elimination}$$

$$\frac{A \wedge B}{B}$$

$(P_{12} \vee P_{21}) \Rightarrow B_{11}$	And elimination
$\neg B_{11} \Rightarrow \neg (P_{12} \vee P_{21})$	Contraposition

$$(A \Rightarrow B) \equiv (\neg B \Rightarrow \neg A)$$

$$\neg(A \lor B) \equiv (\neg A \land \neg B)$$

$\neg B_{11} \Rightarrow \neg (P_{12} \vee P_{21})$	Contraposition
$\neg B_{11} \Rightarrow \neg P_{12} \wedge \neg P_{21}$	"De Morgan"

$B_{11} \Leftrightarrow (P_{12} \vee P_{21})$	Rule of the game
$B_{11} \Rightarrow (P_{12} \vee P_{21}) \wedge (P_{12} \vee P_{21}) \Rightarrow B_{11}$	Biconditional elimination
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$\neg B_{11} \Rightarrow \neg (P_{12} \vee P_{21})$	Contraposition
$\neg B_{11} \Rightarrow \neg P_{12} \wedge \neg P_{21}$	"De Morgan"

Thus, we have <u>proven</u>, in four steps, that no breeze in (1,1) means there can be no pit in either (1,2) or (2,1)

This symbolic inference can be a lot more efficient than naive enumeration of models – if we can apply rules in a good order!

#### The Resolution rule

An inference algorithm is guaranteed to be complete if it uses the *resolution rule* 

$$\frac{A \lor B, \neg B}{A}$$

Unit resolution

$$\frac{A \lor B, \neg B \lor C}{A \lor C}$$

Full resolution

A clause = a disjunction  $(\lor)$  of literals A literal = a positive or a negative symbol

### The Resolution rule

An inference algorithm is guaranteed to be complete if it uses the *resolution rule* 

$$\frac{A_1 \vee A_2 \vee \cdots \vee A_k \vee B, \neg B}{A_1 \vee A_2 \vee \cdots \vee A_k}$$

$$\frac{A_1 \vee A_2 \vee \cdots \vee A_k \vee B, \neg B \vee C_1 \vee C_2 \vee \cdots \vee C_m}{A_1 \vee A_2 \vee \cdots \vee A_k \vee C_1 \vee C_2 \vee \cdots \vee C_m}$$

Note: The resulting clause should only contain one copy of each literal.

### Resolution truth table

Α	В	¬В	С	A∨B	$\neg B \lor C$	A∨C
1	0	1	1	1	1	1
1	1	0	1	1	1	1
0	1	0	1	1	1	1
0	0	1	1	0	1	1
1	0	1	0	1	1	1
1	1	0	0	1	0	1
0	1	0	0	1	0	0
0	0	1	0	0	1	0

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

### Resolution truth table

Α	В	¬В	С	A∨B	$\neg B \lor C$	A∨C
1	0	1	1	1	1	1
1	1	0	1	1	1	1
0	1	0	1	1	1	1
0	0	1	1	0	1	1
1	0	1	0	1	1	1
1	1	0	0	1	0	1
0	1	0	0	1	0	0
0	0	1	0	0	1	0

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

## Conjunctive normal form (CNF)

- Every sentence of propositional logic is equivalent to a conjunction of clauses
  - a clause is a finite disjunction of literals
  - a literal is an atomic formula or its negation
- Sentences expressed in this way are in conjunctive normal form – CNF
  - there is also DNF (disjunctive normal form), i.e.
     a disjunction of conjunctive clauses
- A sentence with exactly k literals per clause is said to be in k-CNF

This is good, it means we can get far with the resolution inference rule.

# Wumpus CNF example

$B_{11} \Leftrightarrow (P_{12} \vee P_{21})$	Rule of the game
$B_{11} \Rightarrow (P_{12} \vee P_{21}) \wedge (P_{12} \vee P_{21}) \Rightarrow B_{11}$	Biconditional elimination
$(\neg B_{11} \lor (P_{12} \lor P_{21})) \land (\neg (P_{12} \lor P_{21}) \lor B_{11})$	Implication elimination
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land ((\neg P_{12} \land \neg P_{21}) \lor B_{11})$	"De Morgan"
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land ((\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21}))$	Distributivity
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land (\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21})$	Voilá – CNF
$(A \Leftrightarrow B) \equiv ((A \Rightarrow B) \land (B \Rightarrow A)) \qquad \neg(A \lor B) \equiv (\neg A)$	∧ ¬B)
$(A \Rightarrow B) \equiv (\neg A \lor B)$ $(A \lor (B \land C)) \equiv$	((A ∨ B) ∧ (A ∨ C))

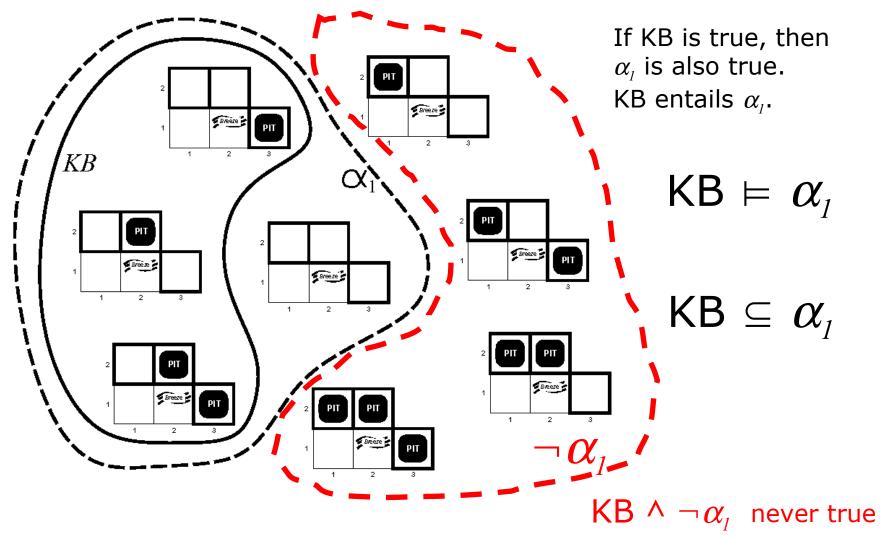
# Wumpus CNF example

$B_{11} \Leftrightarrow (P_{12} \vee P_{21})$	Rule of the game
$B_{11} \Rightarrow (P_{12} \vee P_{21}) \wedge (P_{12} \vee P_{21}) \Rightarrow B_{11}$	Biconditional elimination
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$(\neg B_{11} \lor P_{12} \lor P_{21}) \land (\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21})$	Voilá – CNF

# The **resolution refutation** algorithm

Proves by the principle of contradiction: Shows that  $KB \models \alpha$  by proving that  $(KB \land \neg \alpha)$  is unsatisfiable.

- Convert (KB  $\wedge \neg \alpha$ ) to CNF
- Apply the resolution inference rule repeatedly to the resulting clauses
- Continue until:
  - (a) No more clauses can be added, KB  $\neq \alpha$
  - (b) The empty clause ( $\varnothing$ ) is produced, KB  $\models \alpha$



KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

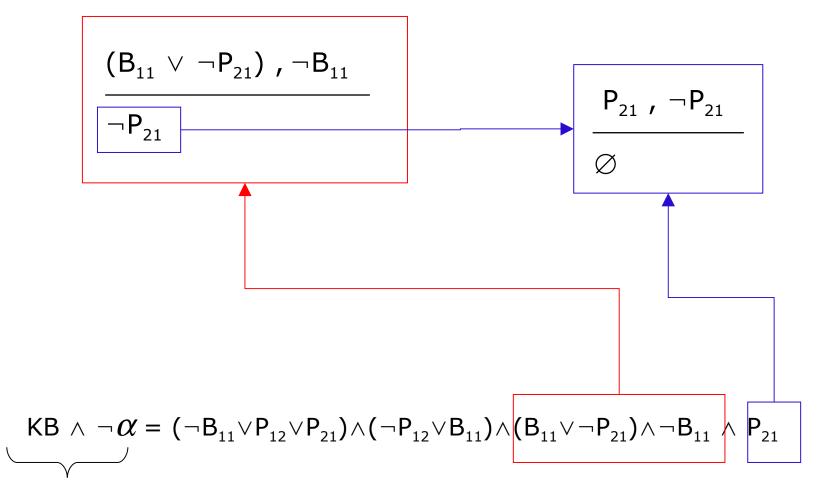
 $\alpha_I$  = The set of models that agrees with conclusion  $\alpha_I$  [conclusion  $\alpha_I$  is true in these models]

### Wumpus resolution example

$B_{11} \Leftrightarrow (P_{12} \vee P_{21})$	Rule of the game
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land (\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21})$	CNF
$\neg B_{11}$	Observation
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land (\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21}) \land \neg B_{11}$	KB in CNF
$\neg P_{21}$	Hypothesis $(\alpha)$

$$\mathsf{KB} \wedge \neg \alpha = (\neg \mathsf{B}_{11} \vee \mathsf{P}_{12} \vee \mathsf{P}_{21}) \wedge (\neg \mathsf{P}_{12} \vee \mathsf{B}_{11}) \wedge (\mathsf{B}_{11} \vee \neg \mathsf{P}_{21}) \wedge \neg \mathsf{B}_{11} \wedge \mathsf{P}_{21}$$

### Wumpus resolution example



Not satisfied, we conclude that  $KB \models \alpha$ 

### Completeness of resolution

S = Set of clauses

RC(S) = Resolution closure of S

RC(S) = Set of all clauses that can be derived from S by the resolution inference rule.

RC(S) has finite cardinality (finite number of symbols  $P_1, P_2, ..., P_k$ )  $\Rightarrow$  Resolution refutation must terminate.

### Completeness of resolution

The ground resolution theorem

If a set S is unsatisfiable, then RC(S) contains the empty clause  $\emptyset$ .

Left without proof.

Your knowledge base (KB) is this:

B

$$B \Rightarrow C$$

$$B \wedge C \Rightarrow A$$

Prove, using the resolution refutation algorithm, that A is True

Your knowledge base (KB) is this:

KB in CNF

$$B \Rightarrow C$$

$$B \wedge C \Rightarrow A$$

B

$$\neg B \lor C$$

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

Prove, using the resolution refutation algorithm, that A is True

Hypothesis: A is True  $\alpha = A$ 

$$KB \wedge \neg \alpha$$

$$B$$

$$\neg B \lor C$$

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

$$\neg A$$

$$KB \wedge \neg \alpha$$

$$B \rightarrow B \lor C \rightarrow (B \land C) \lor A \equiv \neg B \lor \neg C \lor A$$

$$\neg B \lor \neg C \lor A, \neg A$$

$$\neg B \lor \neg C \lor A, \neg A$$

$$B$$

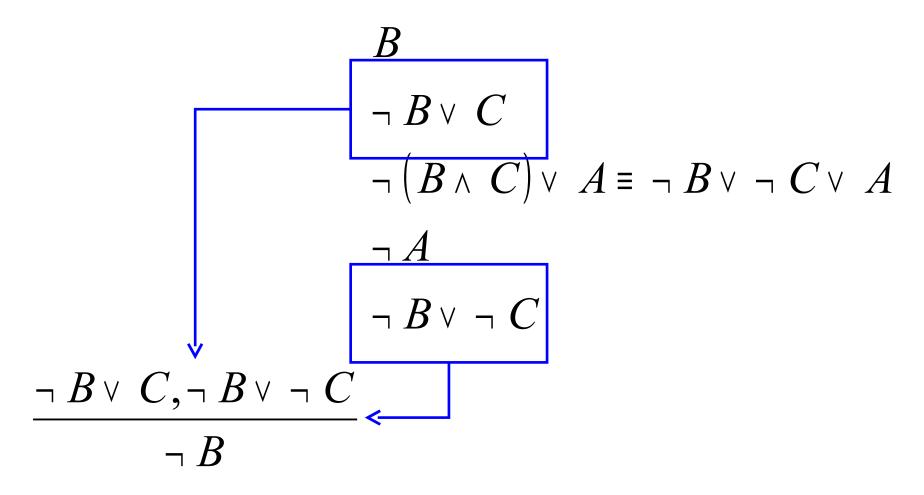
$$\neg B \lor C$$

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

$$\neg A$$

$$\neg B \lor \neg C$$

 $KB \wedge \neg \alpha$ 



$$B$$

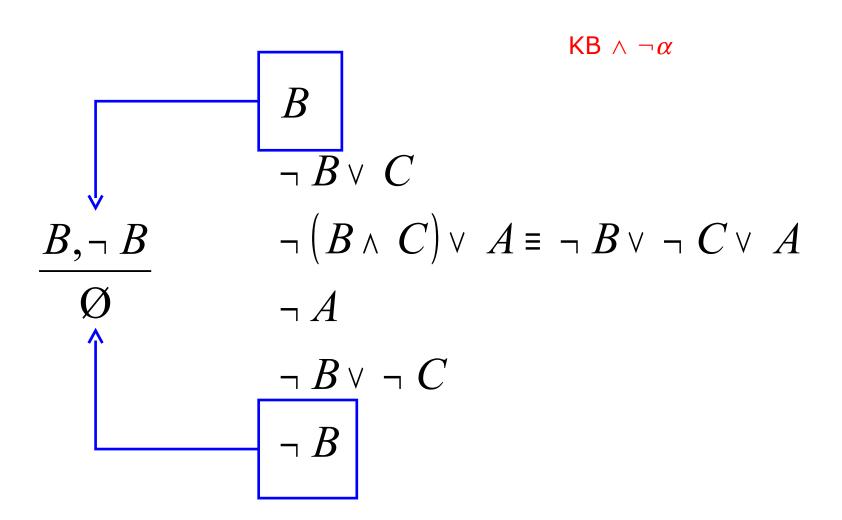
$$\neg B \lor C$$

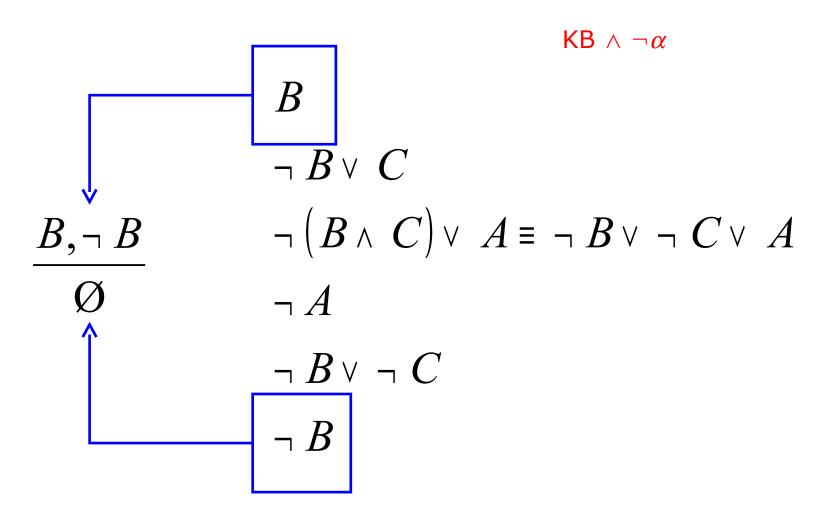
$$\neg (B \land C) \lor A \equiv \neg B \lor \neg C \lor A$$

$$\neg A$$

$$\neg B \lor \neg C$$

$$\neg B$$





(KB  $\wedge \neg \alpha$ ) is unsatisfiable so  $\alpha$  is True.

$$KB \wedge \neg \alpha$$

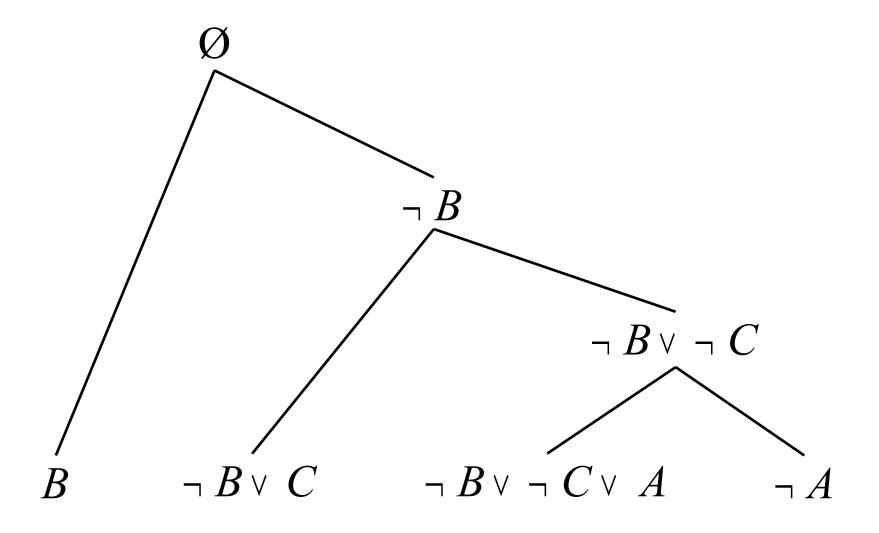
$$B$$

$$\neg B \lor C$$

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

$$\neg A$$

We could have illustrated the resolution refutation steps with a graph...



#### Problem with resolution refutation

- It may expand all nodes (all statements) – exponential in both space and time
- Is there not a more efficient way to only expand those nodes (statements) that affect our query?

# Horn clauses and forward- backward chaining

• Restricted set of clauses: *Horn clauses* disjunction of literals where at most one is positive, e.g.,  $(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k \lor B)$  or  $(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k)$ 

Why Horn clauses?
 Every Horn clause can be written as an implication, e.g.,

$$(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k \lor B) \equiv (A_1 \land A_2 \land \cdots \land A_k) \Rightarrow B$$

$$(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k) \equiv (A_1 \land A_2 \land \cdots \land A_k) \Rightarrow False$$

 Inference in Horn clauses can be done using forward-backward (F-B) chaining in linear time

#### Forward or Backward?

Inference can be run forward or backward

#### Forward-chaining:

Use the current facts in the KB to trigger all possible inferences

#### Backward-chaining:

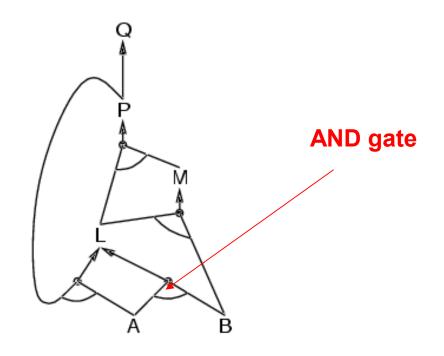
- Work backward from the query proposition Q
- If a rule has Q as a conclusion, see if antecedents can be found to be true

### Example

KB

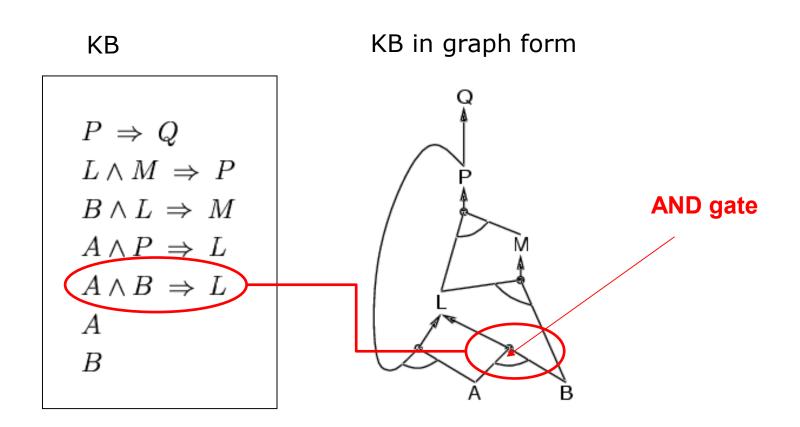
$$P \Rightarrow Q$$
 $L \wedge M \Rightarrow P$ 
 $B \wedge L \Rightarrow M$ 
 $A \wedge P \Rightarrow L$ 
 $A \wedge B \Rightarrow L$ 
 $A$ 

KB in graph form



We are going to check if Q is True

### Example

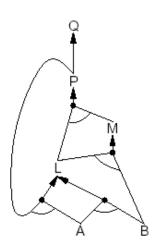


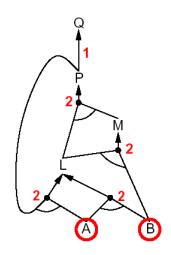
We are going to check if Q is True

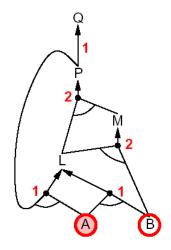
### Example of forward chaining

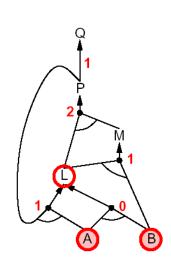
#### We've proved that Q is true

- $P \Rightarrow Q$
- $L \wedge M \Rightarrow P$
- $\begin{array}{c|c}
  B \land L \Rightarrow M \\
  A \land P \Rightarrow L
  \end{array}$
- $A \wedge I \rightarrow L$
- $A \wedge B \Rightarrow L$
- A
- **B**

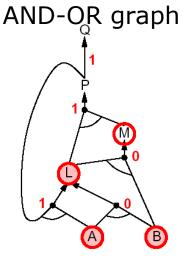


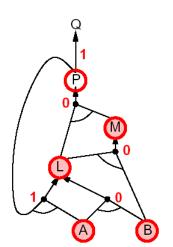


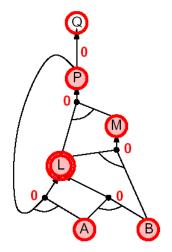




KB







Every step is Modus Ponens, e.g.

$$\frac{A \wedge B \Rightarrow L, A \wedge B}{L}$$

#### Agenda

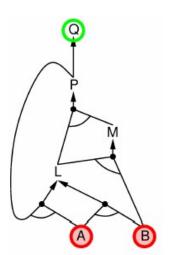
A B L M P Q

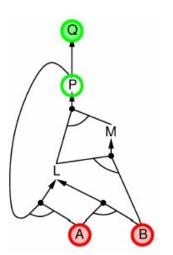
### Example of backward chaining

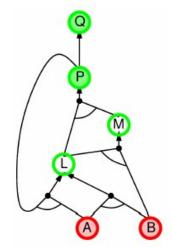
#### Query: is Q true

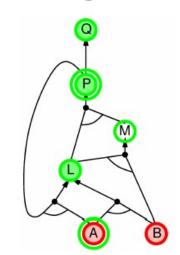
- $P \Rightarrow Q$
- $L \wedge M \Rightarrow P$ 
  - $B \wedge L \Rightarrow M$
- $A \wedge P \Rightarrow L$
- $A \wedge B \Rightarrow L$
- | A
- | B

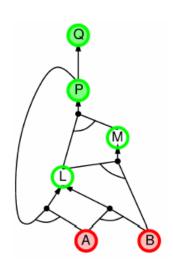
ΚB

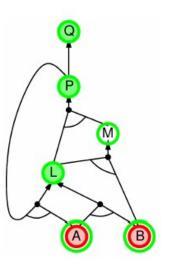


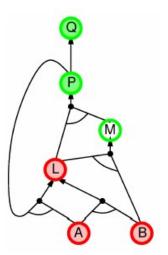


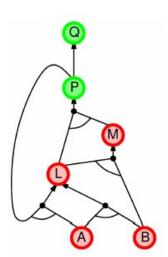


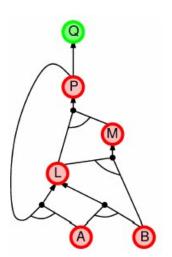






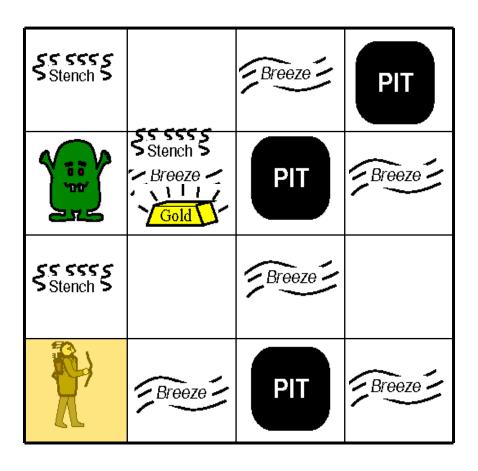






### Wumpus world revisited

Knowledge base (KB) in initial position (ROG = Rule of the Game)



### Wumpus world revisited

Knowledge base (KB) in initial position (ROG = Rule of the Game)

1-16 
$$B_{i,j} \Leftrightarrow (P_{i,j+1} \vee P_{i,j-1} \vee P_{i-1,j} \vee P_{i+1,j})$$

17-32 
$$S_{i,j} \Leftrightarrow (W_{i,j+1} \vee W_{i,j-1} \vee W_{i-1,j} \vee W_{i+1,j})$$

33 
$$(W_{1.1} \vee W_{1.2} \vee W_{1.3} \vee \cdots \vee W_{4.3} \vee W_{4.4})$$

$$34-153 \neg (W_{i,j} \wedge W_{k,l})$$

154 
$$(G_{1,1} \vee G_{1,2} \vee G_{1,3} \vee \cdots \vee G_{4,3} \vee G_{4,4})$$

155-274 
$$\neg (G_{i,j} \land G_{k,l})$$

275 
$$(\neg B_{11} \land \neg W_{11} \land \neg G_{11})$$

ROG: Pits

ROG: Wumpus' odor

ROG: #W ≥ 1

ROG: #W ≤ 1

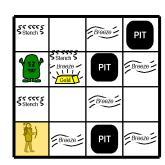
ROG: #G ≥ 1

ROG:  $\#G \le 1$ 

ROG: Start safe

There are 5 "on-states" for every square,  $\{W,P,S,B,G\}$ . A  $4 \times 4$  lattice has  $16 \times 5 = 80$  distinct symbols. Enumerating models means going through  $2^{80}$  models!

The physics rules (1-32) are very unsatisfying – no generalization.



### Summary

- Knowledge is in the form of sentences in a knowledge representation language.
- The representation language has syntax and semantics.
- Propositional logic consists of
  - proposition symbols
  - logical connectives.
- Inference:
  - Model checking
  - Inference rules (e.g. resolution)
    - Horn clauses