## CSE 473 Chapter 7

#### Inference Techniques for Logical Reasoning

 $\neg P_{\scriptscriptstyle 1,2} \lor B_{\scriptscriptstyle 1,1}$ 

 $P_{1,2} \lor P_{2,1} \lor \neg P_{2,1}$ 

 $\neg B_{1,1}$ 

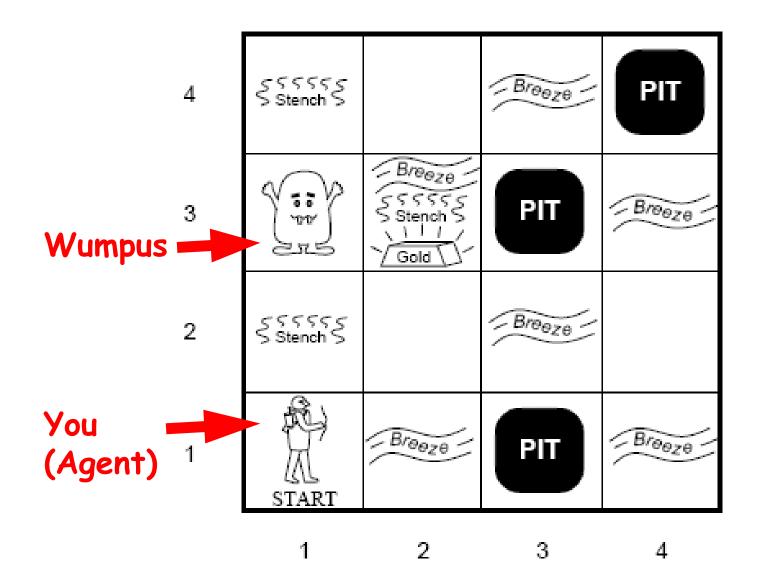
 $\neg P_{2,1}$ 

 $\overline{\neg P_{1,2}}$ 

 $P_{1,2}$ 



## Recall: Wumpus World



## Wumpusitional Logic

Proposition Symbols and Semantics: Let  $P_{i,j}$  be true if there is a pit in [i, j]. Let  $B_{i,j}$  be true if there is a breeze in [i, j].

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2 ок	<sup>2,2</sup> P?	3,2	4,2
1,1 V ОК	<sup>2,1</sup> A B OK	<sup>3,1</sup> P?	4,1

## Wumpus KB

Knowledge Base (KB) includes the following sentences:

 Statements currently known to be true:

• Properties of the world: E.g., "Pits cause breezes in adjacent squares"  $B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1})$ 

 $B_{2,1} \Leftrightarrow (P_{1,1} \lor P_{2,2} \lor P_{3,1})$ (and so on for all squares)

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2	<sup>2,2</sup> P?	3,2	4,2
ок			
1,1 <b>v</b>	<sup>2,1</sup> A B	<sup>3,1</sup> P?	4,1
ок	OK		

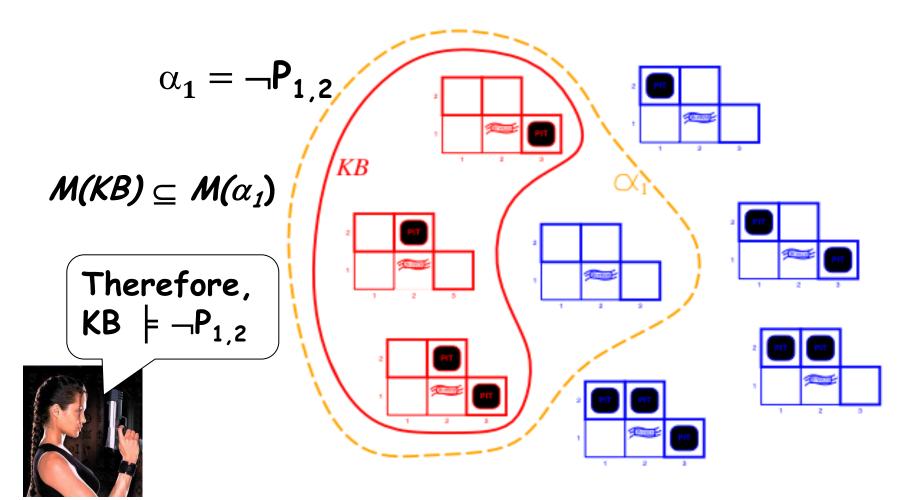


Recall from last time:

*m* is a model of a sentence  $\alpha$  if  $\alpha$  is true in *m* 

 $M(\alpha)$  is the set of <u>all models</u> of  $\alpha$ 

KB  $\models \alpha$  (KB "entails"  $\alpha$ ) iff  $M(KB) \subseteq M(\alpha)$ 



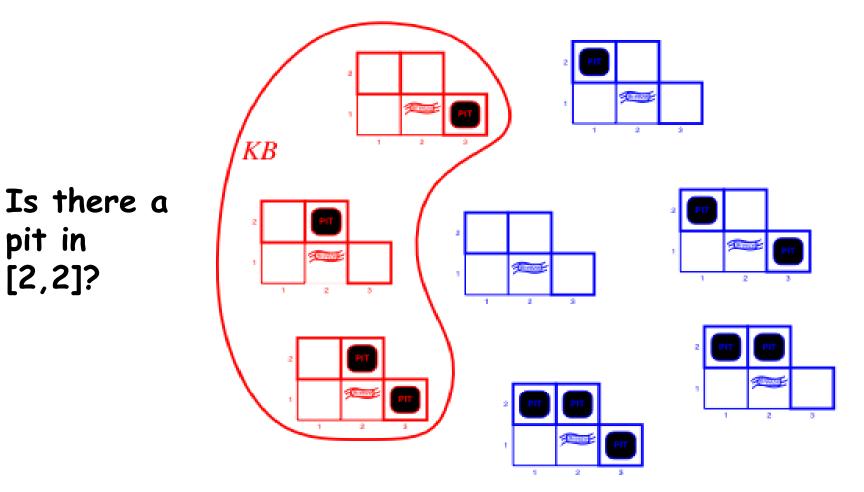
KB = wumpus-world rules + observations

#### Inference by Truth Table Enumeration

	$B_{1,1}$	$B_{2,1}$	P <sub>1,1</sub>	P <sub>1,2</sub>	$P_{2,1}$	$P_{2,2}$	$P_{3,1}$	<b>KB</b>	-P <sub>1,2</sub>
Model 1	false	false	false				false	false	true
Model 2	false	false	false	false	false	false	true	false	true
:	:	:	:	:	:	÷	:	:	:
•	false	true	false	false	false	false	false	false	true
	false	true	false	false	false	false	true	$\underline{true}$	$\underline{true}$
	false	true	false	false	false	true	false	$\underline{true}$	$\underline{true}$
	false	true	false	false	false	true	true	$\underline{true}$	$\underline{true}$
	false	true	false	false	true	false	false	false	true
			:	:	:	:	:	:	:
	true	true	true	true	true	true	true	false	false

In all models in which KB is true,  $\neg P_{1,2}$  is also true Therefore, KB  $\models \neg P_{1,2}$ 

## Another Example



#### Inference by Truth Table Enumeration

$B_{1,1}$	$B_{2,1}$	P <sub>1,1</sub>	$P_{1,2}$	$P_{2,1}$	P <sub>2,2</sub>	$P_{3,1}$	KB
false	false	false	false	false	false	false	false
false	false	false	false	false	false	true	false
:	:	-	:	:	:		:
false	true	false	false	false	false	false	false
false	true	false	false	false	false	true	true
false	true	false	false	false	true	false	<u>true</u>
false	true	false	false	false	true	true	$\underline{true}$
false	true	false	false	true	false	false	false
:	:	:	:	:	:	:	:
true	true	true	true	true	true	true	false

 $P_{2,2}$  is false in a model in which KB is true Therefore, KB  $\not \models P_{2,2}$ 

## Inference by TT Enumeration

- Algorithm: Depth-first enumeration of all models (see Fig. 7.10 in text for pseudocode)
- Algorithm sound?
   Yes
- Algorithm complete?
   Yes
- For *n* symbols, time and space?
- time complexity =  $O(2^n)$ , space = O(n)

#### Other Inference Techniques Rely on Logical Equivalence Laws

Two sentences are logically equivalent iff they are true in the same models:  $a \equiv \beta$  iff  $a \models \beta$  and  $\beta \models a$ 

 $(\alpha \wedge \beta) \equiv (\beta \wedge \alpha)$  commutativity of  $\wedge$  $(\alpha \lor \beta) \equiv (\beta \lor \alpha)$  commutativity of  $\lor$  $((\alpha \land \beta) \land \gamma) \equiv (\alpha \land (\beta \land \gamma))$  associativity of  $\land$  $((\alpha \lor \beta) \lor \gamma) \equiv (\alpha \lor (\beta \lor \gamma))$  associativity of  $\lor$  $\neg(\neg \alpha) \equiv \alpha$  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  $(\alpha \land (\beta \lor \gamma)) \equiv ((\alpha \land \beta) \lor (\alpha \land \gamma))$  distributivity of  $\land$  over  $\lor$  $(\alpha \lor (\beta \land \gamma)) \equiv ((\alpha \lor \beta) \land (\alpha \lor \gamma))$  distributivity of  $\lor$  over  $\land$ 

#### Inference Techniques also rely on Validity and Satisfiability

A sentence is *valid* if it is true in *all* models (a tautology)

e.g., *True*,  $A \lor \neg A$ ,  $A \Rightarrow A$ ,  $(A \land (A \Rightarrow B)) \Rightarrow B$ 

 Validity is connected to inference via the Deduction Theorem:

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

- A sentence is *satisfiable* if it is true in *some* model e.g.,  $A \lor B$ , C
- A sentence is *unsatisfiable* if it is true in no models e.g.,  $A \land \neg A$
- Satisfiability is connected to inference via the following:  $KB \models a$  if and only if  $(KB \land \neg a)$  is unsatisfiable (proof by contradiction)

## Inference/Proof Techniques

- Two kinds (roughly):
  - 1. Model checking
    - Truth table enumeration (always exponential in n)
    - Efficient backtracking algorithms,
      - e.g., Davis-Putnam-Logemann-Loveland (DPLL)
    - Local search algorithms (sound but incomplete) e.g., randomized hill-climbing (WalkSAT)
  - 2. Successive application of inference rules
    - Generate new sentences from old in a sound way
    - Proof = a sequence of inference rule applications
    - Use inference rules as successor function in a standard search algorithm

Let us look at a #2 type technique: Resolution...

### Inference Technique I: <u>Resolution</u> Motivation

There is a pit in [1,3] or There is a pit in [2,2]

There is no pit in [2,2]

There is a pit in [1,3]

More generally,  $\frac{\ell_{1} \vee ... \vee \ell_{k}, \qquad \neg \ell_{i}}{\ell_{1} \vee ... \vee \ell_{i-1} \vee \ell_{i+1} \vee ... \vee \ell_{k}}$ 



Terminology:

- Literal = proposition symbol or its negation E.g., A,  $\neg A$ , B,  $\neg B$ , etc.
- Clause = disjunction of literals E.g.,  $(B \lor \neg C \lor \neg D)$

Resolution assumes sentences are in Conjunctive Normal Form (CNF): sentence = conjunction of clauses E.g.,  $(A \lor \neg B) \land (B \lor \neg C \lor \neg D)$ 

### Conversion to CNF

 $\mathsf{E.g., B}_{1,1} \iff (\mathsf{P}_{1,2} \lor \mathsf{P}_{2,1})$ 

- 1. Eliminate  $\Leftrightarrow$ , replacing  $a \Leftrightarrow \beta$  with  $(a \Rightarrow \beta) \land (\beta \Rightarrow a)$ .  $(B_{1,1} \Rightarrow (P_{1,2} \lor P_{2,1})) \land ((P_{1,2} \lor P_{2,1}) \Rightarrow B_{1,1})$
- 2. Eliminate  $\Rightarrow$ , replacing  $a \Rightarrow \beta$  with  $\neg a \lor \beta$ .  $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg (P_{1,2} \lor P_{2,1}) \lor B_{1,1})$
- 3. Move  $\neg$  inwards using de Morgan's rule: ( $\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}$ )  $\land$  (( $\neg P_{1,2} \land \neg P_{2,1}$ )  $\lor B_{1,1}$ )
- 4. Apply distributivity law ( $\land$  over  $\lor$ ) and flatten: ( $\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}$ )  $\land$  ( $\neg P_{1,2} \lor B_{1,1}$ )  $\land$  ( $\neg P_{2,1} \lor B_{1,1}$ )

This is in CNF - Done!

#### Inference Technique: Resolution

• General Resolution inference rule (for CNF):  $\frac{l_{1} \vee ... \vee l_{i} ... \vee l_{k}, \quad m_{1} \vee ... \vee m_{j} ... \vee m_{n}}{l_{1} \vee ... \vee l_{i-1} \vee l_{i+1} \vee ... \vee l_{k} \vee m_{1} \vee ... \vee m_{j-1} \vee m_{j+1} \vee ... \vee m_{n}}$ where  $l_{i}$  and  $m_{j}$  are complementary literals i.e.  $l_{i} = \neg m_{j}$ .

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

<b>P</b> <sup>•?</sup>		
в ок А А	<u>&gt;</u>	
ок    А	s ок —>А	

#### Soundness of Resolution Inference Rule

(Recall logical equivalence  $A \Rightarrow B \equiv \neg A \lor B$ ) Express each clause as:

$$\neg (l_1 \vee ... \vee l_{i-1} \vee l_{i+1} \vee ... \vee l_k) \Rightarrow l_i$$
  
$$\neg m_j \Rightarrow (m_1 \vee ... \vee m_{j-1} \vee m_{j+1} \vee ... \vee m_n)$$

 $\neg(l_{i} \vee ... \vee l_{i-1} \vee l_{i+1} \vee ... \vee l_{k}) \Rightarrow (m_{1} \vee ... \vee m_{j-1} \vee m_{j+1} \vee ... \vee m_{n})$ (since  $l_{i} = \neg m_{j}$ )

## Resolution algorithm

To show KB = a, use proof by contradiction,
 i.e., show KB ^ ¬a unsatisfiable

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function PL-RESOLUTION(KB, \alpha) returns true or false

clauses \leftarrow the set of clauses in the CNF representation of KB \land \neg \alpha

new \leftarrow \{\}

loop do

for each C_i, C_j in clauses do

resolvents \leftarrow PL-RESOLVE(C_i, C_j)

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
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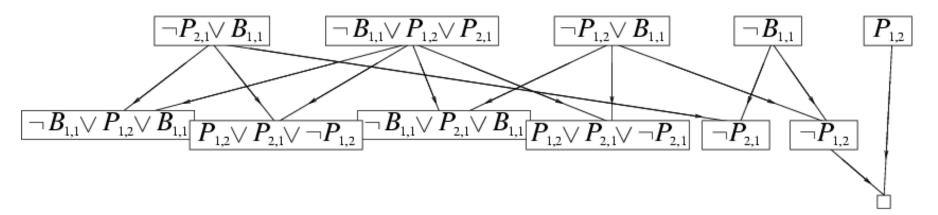
## **Resolution** example

Given no breeze in [1,1], prove there's no pit in [1,2]

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

Resolution: Convert to CNF and show KB  $\wedge \neg \alpha$  is unsatisfiable

## **Resolution** example



# (i.e., KB $\land \neg$ a unsatisfiable)

## Next Time

- WalkSAT
- Logical Agents: Wumpus
- First-Order Logic
- To Do: Project #2 Finish Chapter 7 Start Chapter 8