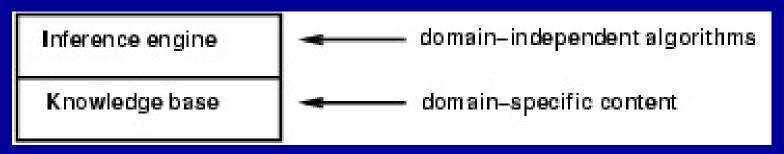
# Chapter 7 Logical Agents

#### Outline

- Knowledge-based agents
- Wumpus world
- Logic in general models and entailment
- Propositional (Boolean) logic
- Equivalence, validity, satisfiability
- Inference rules and theorem proving

# Knowledge bases



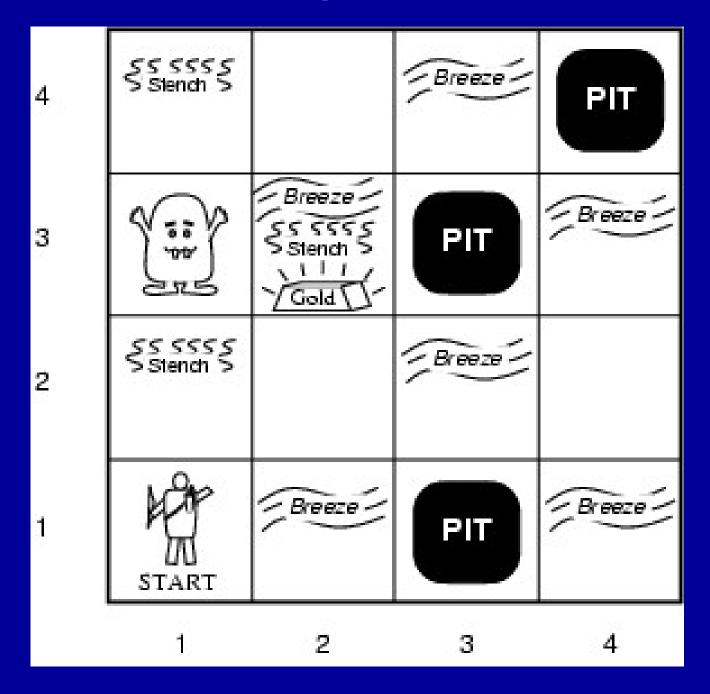
- ☐ The central component of a knowledge-based agent is its knowledge base, or KB.
- ☐ Knowledge base(KB) = set of sentences in a formal language
- ☐ Each time the agent program is called, it does three things.
  - 1) it TELLs the knowledge base what it perceives.
  - 2) it ASKs the knowledge base what action it should perform.
  - 3) the agent program TELLs the knowledge base which action was chosen, and the agent executes the action.

#### A simple knowledge-based agent

#### ☐ The agent must be able to:

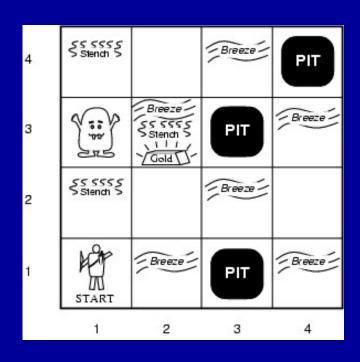
- Represent states, actions, etc.
- Incorporate new percepts
- Update internal representations of the world
- Deduce hidden properties of the world
- Deduce appropriate actions

# A Wumpus World



#### Wumpus World PEAS description

- □ Performance measure
  - gold +1000, death -1000
  - -1 per step, -10 for using the arrow
- ☐ Environment: 4 x 4 grid of rooms
  - Squares adjacent to wumpus are smelly
  - Squares adjacent to pit are breezy
  - Glitter iff 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



- Sensors: Stench, Breeze, Glitter, Bump, Scream (shot Wumpus)
- Actuators: Left turn, Right turn, Forward, Grab, Release, Shoot

- □Fully Observable
- □ Deterministic
- □ <u>Episodic</u>
- □Static
- □ Discrete
- ☐ Single-agent?

- □Fully Observable No only local perception
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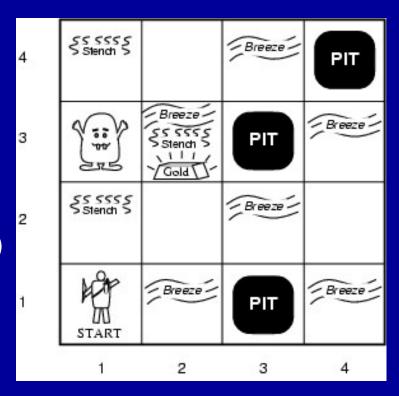
- □Fully Observable No only local perception
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- □Fully Observable No only local perception
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- □ Static Yes Wumpus and Pits do not move
- □ <u>Discrete</u> Yes
- □ Single-agent? Yes Wumpus is essentially a natural feature

#### Wumpus World

- Percepts given to the agent
- 1. Stench
- 2. Breeze
- 3. Glitter/Gold
- 4. Bumb (ran into a wall)
- 5. Scream (wumpus has been hit by arrow)



☐ Principle Difficulty: agent is initially ignorant of the configuration of the environment – going to have to reason to figure out where the gold is without getting killed!

# **Exploring the Wumpus World**

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 OK	2,1 OK	3,1	4,1

(a)

A = Agent

 $\overline{\mathbf{B}}$  = Breeze

G = Glitter, Gold

OK = Safe square

P = Pit

S = Stench

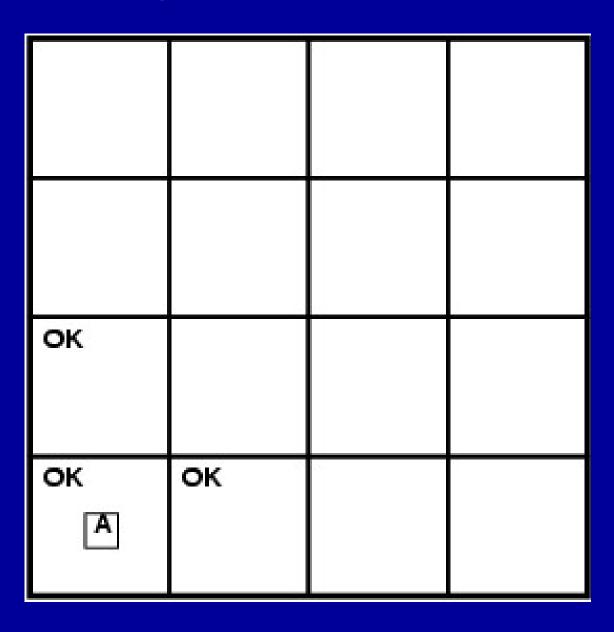
V = Visited

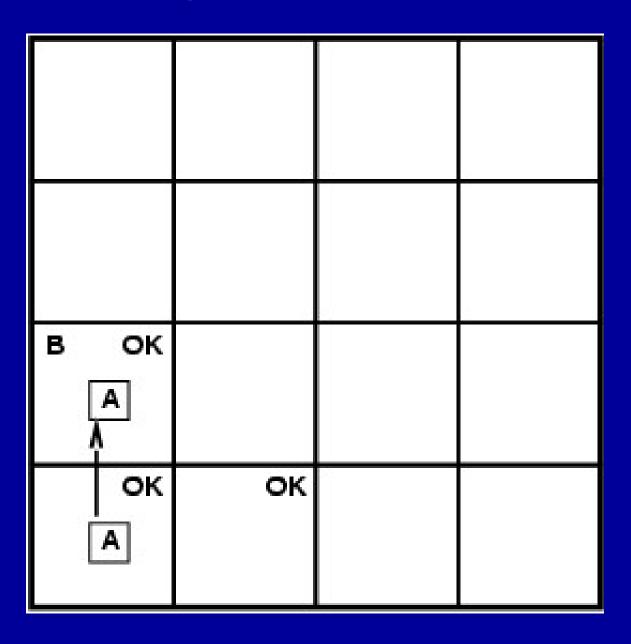
W = Wumpus

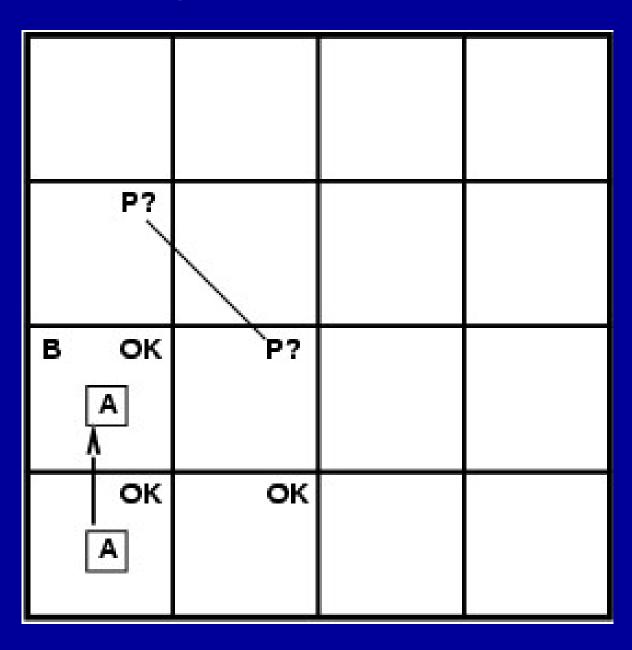
#### **Initial situation:**

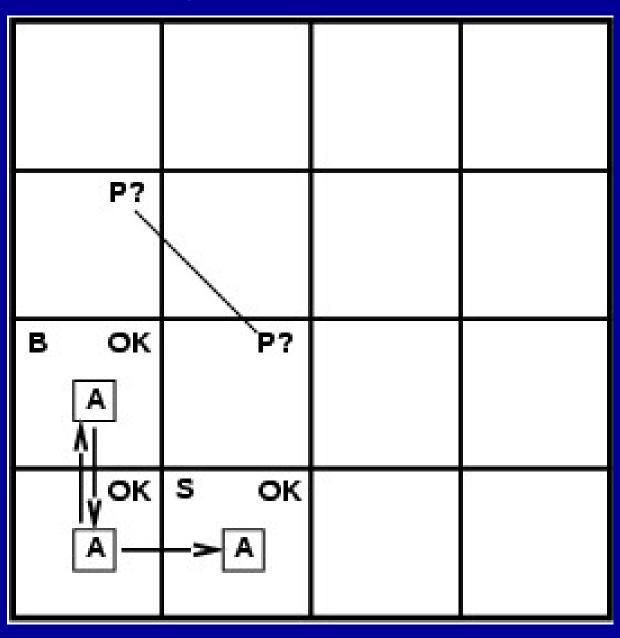
Agent in 1,1 and percept is [None, None, None, None]

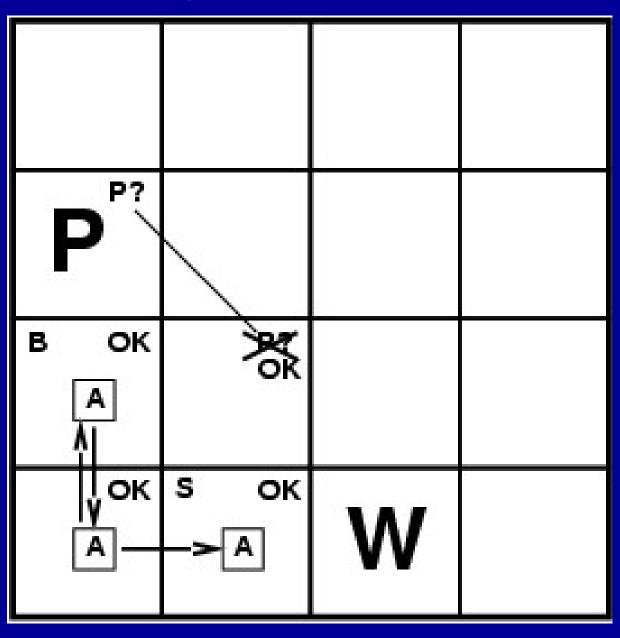
From this the agent can infer the neighboring squares are safe (otherwise there would be a breeze or a stench)

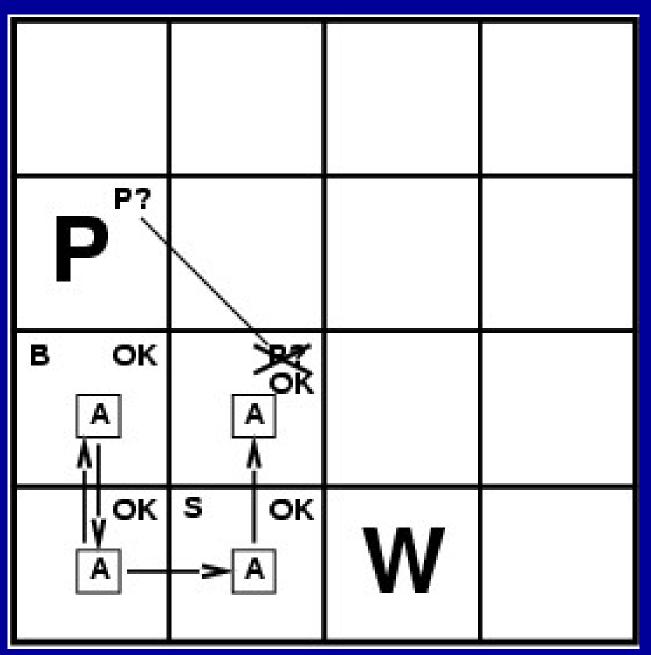


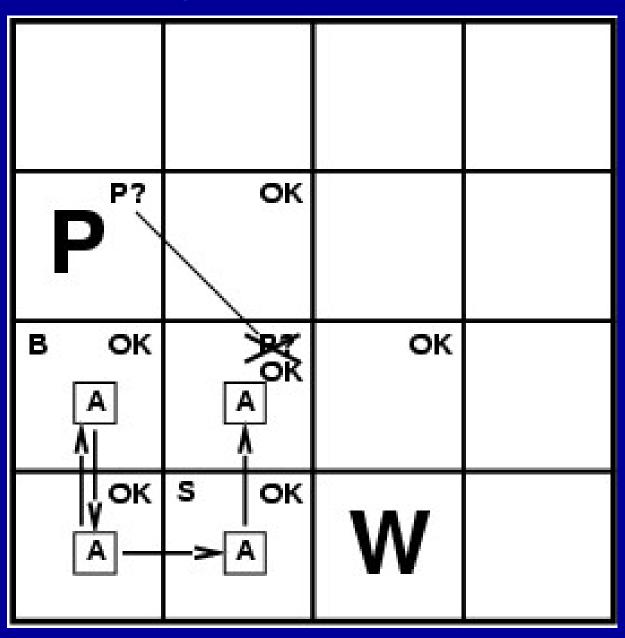


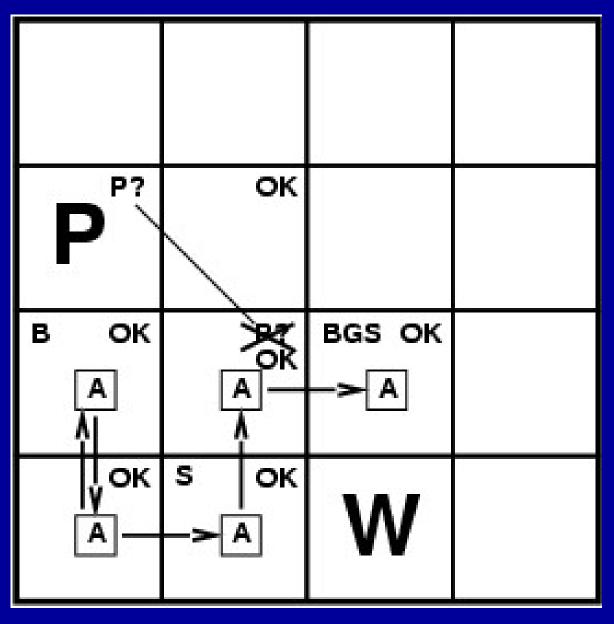


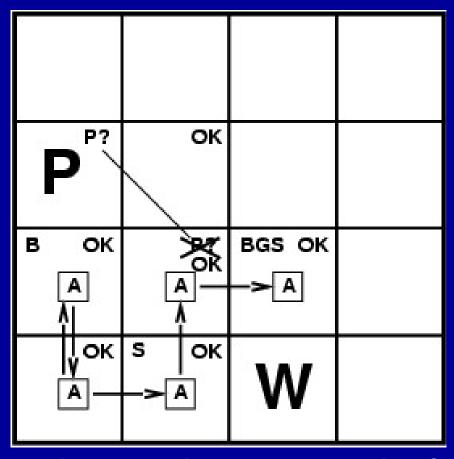












In each case where the agent draws a conclusion from the available Information, that conclusion is guaranteed to be correct if the available Information is correct...

This is a fundamental property of logical reasoning

# Logic in general

- Logics are formal languages for representing information such that conclusions can be drawn
- Syntax defines how symbos can be put together to form the sentences in the language
- ☐ Semantics define the "meaning" of sentences;
  - i.e., define truth of a sentence in a world (given an interpretation)
- E.g., the language of arithmetic
  - x+2 ≥ y is a sentence; x2+y > {} is not a sentence
  - $-x+2 \ge y$  is true iff the number x+2 is no less than the number y
  - $x+2 \ge y$  is true in a world where x = 7, y = 1
  - $x+2 \ge y$  is false in a world where x = 0, y = 6

#### Entailment

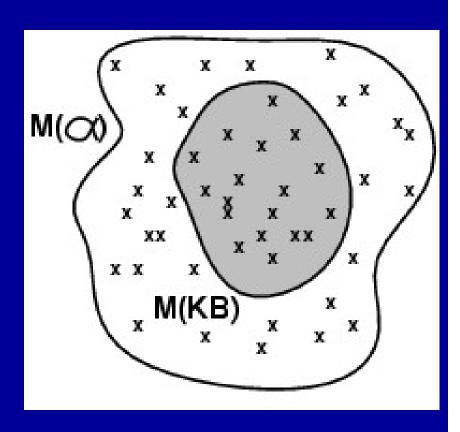
 Entailment means that one thing follows logically from another:

- Knowledge base KB entails sentence α if and only if α is true in all worlds where KB is true
  - E.g., the KB containing "the Phillies won" and "the Reds won" entails "Either the Phillies won or the Reds won"
  - E.g., x+y = 4 entails 4 = x+y
  - Entailment is a relationship between sentences (i.e., syntax) that is based on semantics

#### Models

- ☐ Logicians typically think in terms of models, which are formally structured worlds with respect to which truth can be evaluated
- $\square$  We say m is a model of a sentence  $\alpha$  if  $\alpha$  is true in m
- $\square$   $M(\alpha)$  is the set of all models of  $\alpha$
- $\square$  Then KB  $\models \alpha$  iff  $M(KB) \subseteq M(\alpha)$

E.g. KB = Phillies won and Yankees won  $\alpha$  = Phillies won

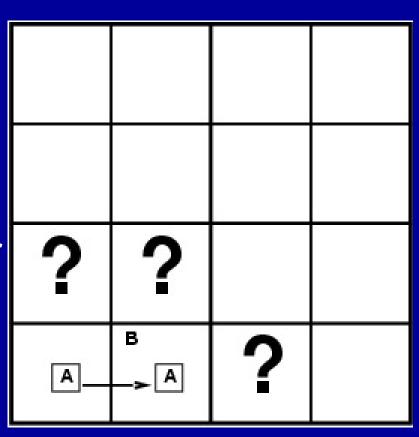


### Entailment in the wumpus world

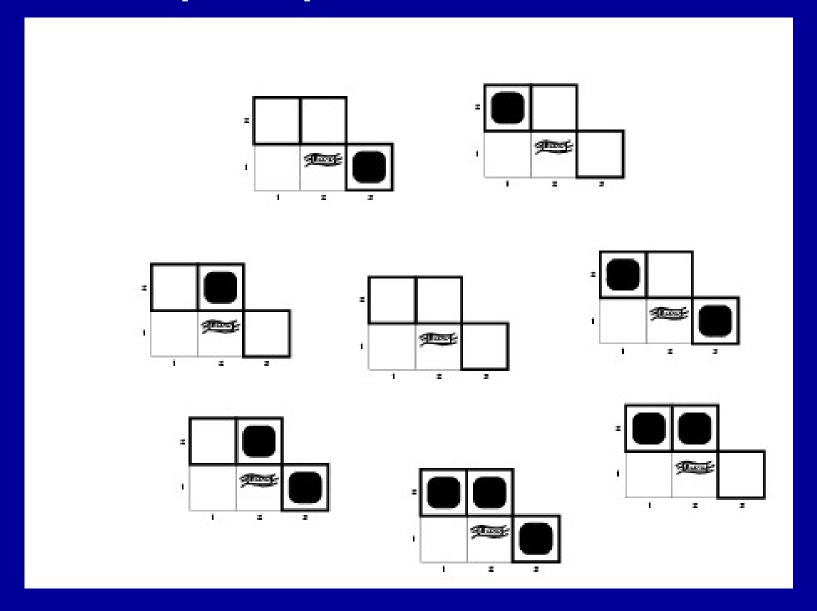
Situation after detecting nothing in [1,1], moving right, breeze in [2,1]

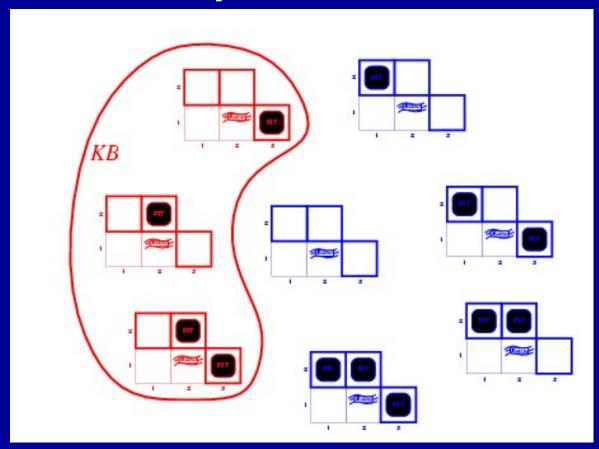
Consider possible models for KB assuming only pits

3 Boolean choices ⇒ 8 possible models

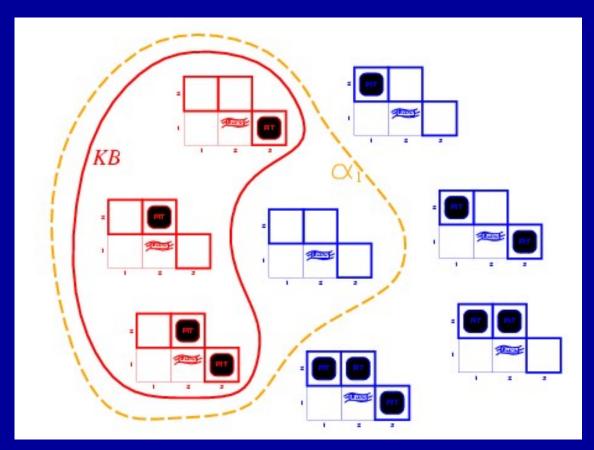


### Wumpus possible models

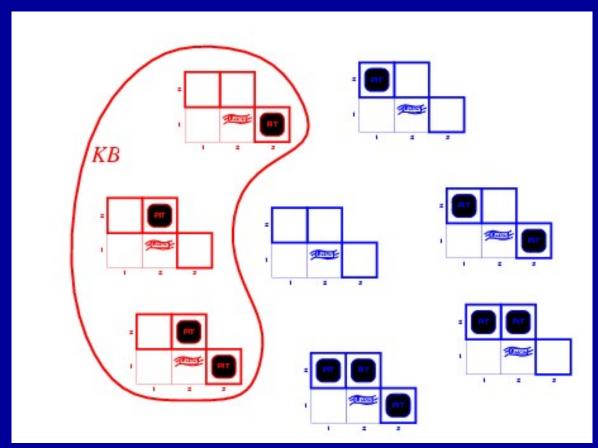




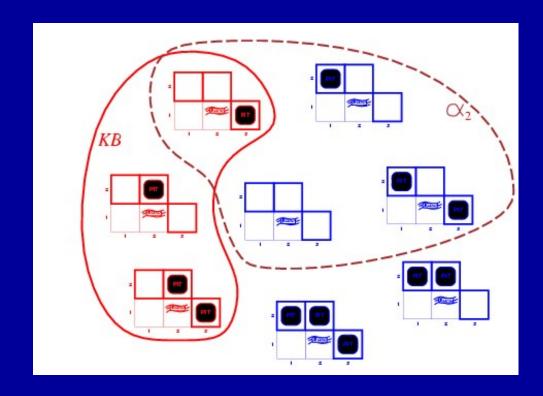
KB = wumpus-world rules + observations



- KB = wumpus-world rules + observations
- $\alpha_1$  = "there is no pit in [1,2]",  $KB \models \alpha_1$ , proved by model checking



• *KB* = wumpus-world rules + observations



- KB = wumpus-world rules + observations
- $\alpha_2$  = "there is no pit in [2/,2]",  $KB = \alpha_2$

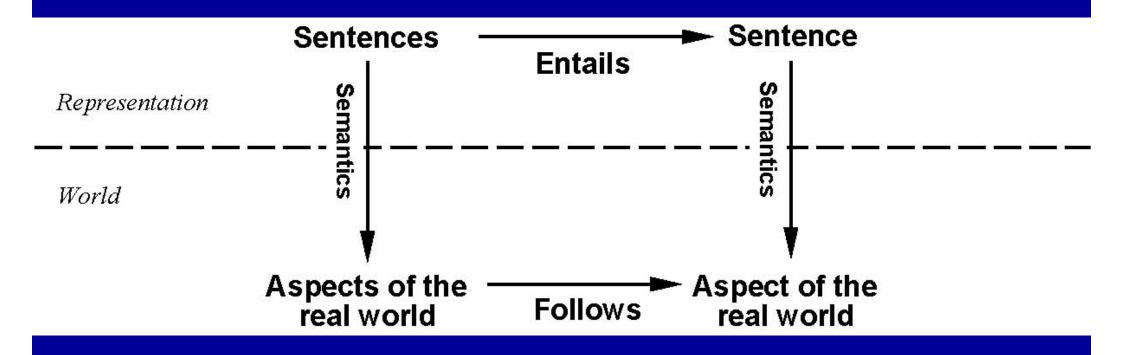
#### Inference and Entailment

- Inference is a procedure that allows new sentences to be derived from a knowledge base.
- Understanding inference and entailment: think of
  - Set of all consequences of a KB as a haystack
  - α as the needle
- Entailment is like the needle being in the haystack
- Inference is like finding it

#### Inference

- KB | α = sentence α can be derived from KB by inference procedure I
- Soundness: *i* is sound if whenever  $KB \mid_{i} \alpha$ , it is also true that  $KB \models \alpha$
- Completeness: *i* is complete if whenever  $KB = \alpha$ , it is also true that  $KB = \alpha$
- Preview: we will define a logic (first-order logic) which is expressive enough to say almost anything of interest, and for which there exists a sound and complete inference procedure.
- That is, the procedure will answer any question whose answer follows from what is known by the *KB*.

#### Step Back...



This is an inference procedure whose conclusions are guaranteed to be true In any world where the premises are true.

If KB is true in the real world, then any sentence  $\alpha$  derived from KB by a sound inference procedure is also true in the real world.

## Propositional logic: Syntax

- Propositional logic is the simplest logic illustrates basic ideas
- The proposition symbols P<sub>1</sub>, P<sub>2</sub> etc are (atomic) sentences
  - If S is a sentence, ¬(S) is a sentence (negation)
  - If  $S_1$  and  $S_2$  are sentences,  $(S_1 \wedge S_2)$  is a sentence (conjunction)
  - If  $S_1$  and  $S_2$  are sentences,  $(S_1 \vee S_2)$  is a sentence (disjunction)
  - If  $S_1$  and  $S_2$  are sentences,  $(S_1 \Rightarrow S_2)$  is a sentence (implication)
  - If  $S_1$  and  $S_2$  are sentences,  $(S_1 \Leftrightarrow S_2)$  is a sentence (biconditional)

## Propositional logic: Semantics

Each model specifies true/false for each proposition symbol

E.g. 
$$P_{1,2}$$
  $P_{2,2}$   $P_{3,1}$  false true false

With these symbols, 8 possible models, can be enumerated automatically.

Rules for evaluating truth with respect to a model *m*:

¬S	is true iff	S is false	
$S_1 \wedge S_2$	is true iff	S <sub>1</sub> is true and	S <sub>2</sub> is true
$S_1 \vee S_2$	is true iff	S₁is true or	S <sub>2</sub> is true
$S_1 \Rightarrow S$	2 is true iff	S <sub>1</sub> is false or	S <sub>2</sub> is true
i.e.,	is false iff	S <sub>1</sub> is true and	S <sub>2</sub> is false
$S_1 \Leftrightarrow S$	<sub>2</sub> is true iff	$S_1 \Rightarrow S_2$ is true a	ndS <sub>2</sub> ⇒S <sub>1</sub> is true

Simple recursive process evaluates an arbitrary sentence, e.g.,

$$\neg P_{1,2} \land (P_{2,2} \lor P_{3,1}) = true \land (true \lor false) = true \land true = true$$

### Truth tables for connectives

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

### Truth tables for connectives

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

John likes football and John likes baseball. John likes football or John likes baseball. (English or is a bit different...)

### Truth tables for connectives

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

John likes football and John likes baseball.

John likes football or John likes baseball.

If John likes football then John likes baseball.

(Note different from English – if John likes football maps to false, then the sentence is true.)

(Implication seems to be if antecedent is true then I claim the consequence is, otherwise I make no claim.)

## Wumpus world sentences

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].

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

"Pits cause breezes in adjacent squares"

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

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

### Simple Inference Procedure

- $KB \models \alpha$ ?
- Model checking enumerate the models, and check if α is true in every model in which KB is true. Size of truth table depends on # of atomic symbols.
- Remember a model is a mapping of all atomic symbols to true or false – use semantics of connectives to come to an interpretation for them.

### Truth tables for inference

$B_{1,1}$	$B_{2,1}$	$P_{1,1}$	$P_{1,2}$	$P_{2,1}$	$P_{2,2}$	$P_{3,1}$	KB	$\alpha_1$
false	true							
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	false	false						

## Inference by enumeration

Depth-first enumeration of all models is sound and complete

```
function TT-Entails?(KB, \alpha) returns true or false
symbols \leftarrow \text{a list of the proposition symbols in } KB \text{ and } \alpha
\text{return TT-Check-All}(KB, \alpha, symbols, [])
function TT-Check-All}(KB, \alpha, symbols, model) \text{ returns } true \text{ or } false
\text{if Empty?}(symbols) \text{ then}
\text{if PL-True?}(KB, model) \text{ then return PL-True?}(\alpha, model)
\text{else return } true
\text{else do}
P \leftarrow \text{First}(symbols); rest \leftarrow \text{Rest}(symbols)
\text{return TT-Check-All}(KB, \alpha, rest, \text{Extend}(P, true, model) \text{ and}
\text{TT-Check-All}(KB, \alpha, rest, \text{Extend}(P, false, model)
```

• For *n* symbols, time complexity is  $O(2^n)$ , space complexity is O(n)

## Logical equivalence

□ Two sentences are logically equivalent iff true in same models:  $\alpha \equiv \beta$  iff  $\alpha \models \beta$  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 and satisfiability

A sentence is valid if it is true in all models, 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:  $KB \models \alpha$  if and only if  $(KB \Rightarrow \alpha)$  is valid

A sentence is satisfiable if it is true in some model e.g., Av B, C

A sentence is unsatisfiable if it is true in no models e.g., A \ A

Satisfiability is connected to inference via the following:  $KB \models \alpha$  if and only if  $(KB \land \neg \alpha)$  is unsatisfiable

### 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 algorithm
    - Typically require transformation of sentences into a normal form
  - Model checking
    - truth table enumeration (always exponential in n)
    - improved backtracking, e.g., Davis--Putnam-Logemann-Loveland (DPLL)
    - heuristic search in model space (sound but incomplete)
       e.g., min-conflicts-like hill-climbing algorithms

### Conversion to CNF

- $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$
- 1. Eliminate  $\Leftrightarrow$ , replacing  $\alpha \Leftrightarrow \beta$  with  $(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$ .  $(B_{1,1} \Rightarrow (P_{1,2} \lor P_{2,1})) \land ((P_{1,2} \lor P_{2,1}) \Rightarrow B_{1,1})$
- 2. Eliminate  $\Rightarrow$ , replacing  $\alpha \Rightarrow \beta$  with  $\neg \alpha \lor \beta$ .  $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg (P_{1,2} \lor P_{2,1}) \lor B_{1,1})$
- 3. Move 

  inwards using de Morgan's rules and double-negation:

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

4. Apply distributivity law ( $\land$  over  $\lor$ ) and flatten:  $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg P_{1,2} \lor B_{1,1}) \land (\neg P_{2,1} \lor B_{1,1})$ 

### Resolution example

•  $KB = (B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})) \land \neg B_{1,1} \alpha = \neg P_{1,2}$ 

