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# Using Logic for Knowledge Bases

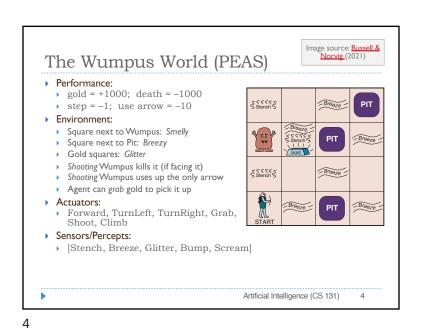
- ▶ A logic is a formal language:
  - ▶ Represents facts that we know about the world: each such fact is a sentence in our logical language
  - Allows us to generate new facts based on what we already know by performing inference
- ▶ A logic has two basic components:
  - ▶ Syntax: defines what counts as a sentence of the language
  - ▶ Semantics: defines what makes a sentence true or false

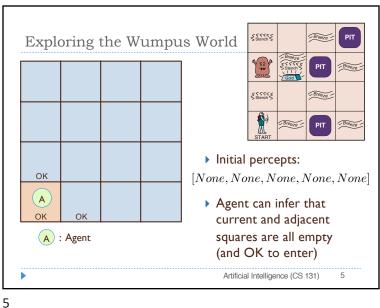
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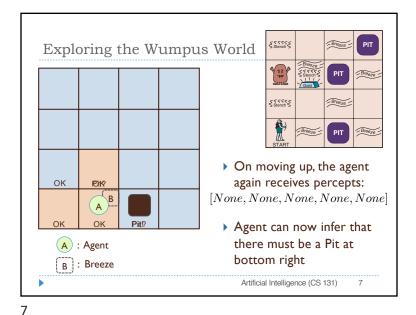
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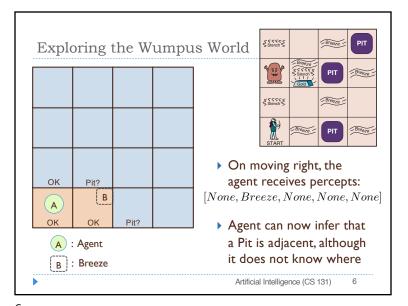
Image source: Russell & **Knowledge-Based Agents** We can build AI agents that act in the world based upon incoming data, supplemented by their knowledge of the world Actions are function KB-AGENT(percept) returns an action persistent: KB, a knowledge base chosen based t, a counter, initially 0, indicating time upon *KB*, a knowledge-Tell(KB, Make-Percept-Sentence(percept, t))base  $action \leftarrow Ask(KB, Make-Action-Query(t))$ Tell(KB, Make-Action-Sentence(action, t)) $t \leftarrow t + 1$ return action Incoming data (percepts) and chosen actions are also recorded to that knowledge-base Artificial Intelligence (CS 131)

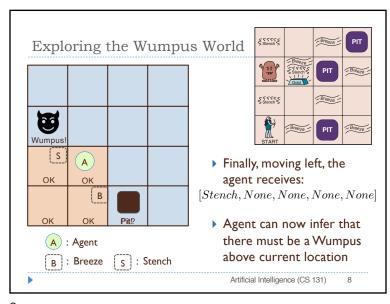
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#### Inference & Logical Entailment

- An inference rule tells an agent how to draw conclusions based upon what they already know
- ▶ Valid rules encode a semantic relationship between what we infer and and what is already in our knowledge base
- We say that a sentence  $\alpha$  follows from our knowledge base, KB, or that KB entails  $\alpha$ :

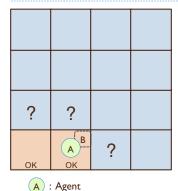
 $KB \models \alpha \Leftrightarrow \alpha$  is true in every situation in which KB is true

▶ For example:

 $\{\text{Breeze at } [2,1]\} \models (\text{Pit at } [3,1]) \text{ OR } (\text{Pit at } [2,2])$ 

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B: Breeze

▶ The agent starts at bottom left, sensing nothing adjacent to it:

[None, None, None, None, None]

On moving right, the agent senses a breeze:

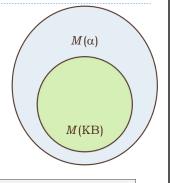
[None, Breeze, None, None, None]

▶ Based on this sequence of percepts, there exist a total of 8 possible models representing the presence or absence of pits in each of the 3 adjacent squares

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#### Semantic Models

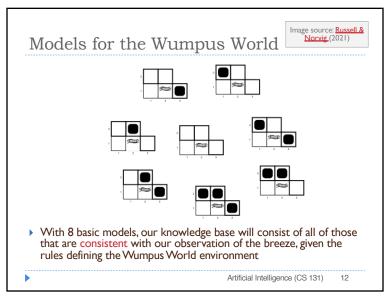
- A formal semantics defines truth and entailment in terms of a logical model
- Circumstances under which a sentence is true or false
- Model m is a model of sentence  $\alpha$  if  $\alpha$  is true in m
- $M(\alpha)$  = set of all models of  $\alpha$
- $\blacktriangleright$  KB entails  $\alpha$  whenever all models of KB are also models of  $\alpha$

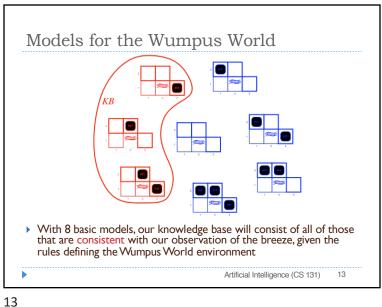


 $KB \models \alpha \Leftrightarrow M(KB) \subseteq M(\alpha)$ 

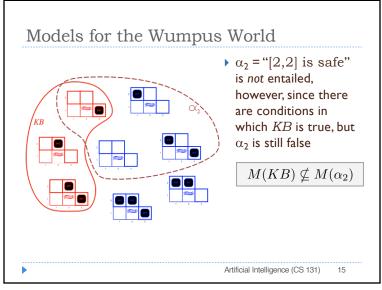
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Models for the Wumpus World  $\alpha_1=\text{``[1,2] is safe''} \text{ is entailed by our knowledge base, due to the containment relationship between the models}$   $M(KB)\subseteq M(\alpha_1)$  Artificial Intelligence (CS 131) 14

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## Syntax of Propositional Logic (PL)

- ▶ A basic formal language for representing simple statements of fact (propositions)
- A set of logical connectives and a simple recursive syntax (grammar) for their combination into sentences
- Basic proposition symbols,  $P_1, P_2, ...,$  are sentences
- If S is a sentence,  $\neg S$  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)

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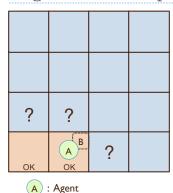
# Semantics of Propositional Logic (PL)

- A PL model m is an assignment of true/false values to whatever basic propositional symbols exist
  - $P_1$  = True,  $P_2$  = False,  $P_3$  = True, etc.
  - $\blacktriangleright$  For *n* propositional symbols,  $2^n$  such models are possible
- ▶ Given an assignment to basic symbols, all more complex sentences are true/false according to semantic rules:
- $\neg S$  is true iff (if and only if) S is false
- $S_1 \wedge S_2$  is true iff  $S_1$  is true and  $S_2$  is true
- $S_1 \vee S_2$  is true iff  $S_1$  is true or  $S_2$  is true
- $S_1 \Rightarrow S_2$  is true iff  $S_1$  is false or  $S_2$  is true
- $S_1 \Leftrightarrow S_2$  is true iff  $S_1 \Rightarrow S_2$  is true and  $S_2 \Rightarrow S_1$  is true

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# Logic in the Wumpus World



B: Breeze

- We can use PL to express the situation of the agent:
- Let  $P_{i,i}$  be true if there is a pit in location [i, j]
- Let  $B_{i,i}$  be true if there is a breeze in location [i, j]
- · The relevant KB is:

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

▶ PL can also express Wumpus World rules like "Pits cause breezes in adjacent squares":

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

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

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#### Truth Tables for PL Semantics

P	Q	$\neg P$	$P \wedge Q$	$P \lor Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
Т	Т	F	Т	Т	Т	Т
T	F	F	F	Т	F	F
F	Т	Т	F	Т	Т	F
F	F	Т	F	F	Т	Т

- The basic semantic rules of PL correspond to a simple set of truth tables, each of which gives the result of applying one connective to one or more propositional symbols
- Each connective is defined by its truth function: takes one or more truth values as input and outputs another truth value

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# Truth-Functional Analysis (01)

▶ Based on the basic truth tables, we can assign a semantic value to any complex PL sentence, recursively

 $P \quad Q \qquad \neg (P \lor Q) \Leftrightarrow (\neg P \land \neg Q)$ 

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## Truth-Functional Analysis (02)

▶ Based on the basic truth tables, we can assign a semantic value to any complex PL sentence, recursively

$$P \quad Q \qquad \neg (P \vee Q) \Leftrightarrow (\neg P \wedge \neg Q)$$

FF

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## Truth-Functional Analysis (04)

▶ Based on the basic truth tables, we can assign a semantic value to any complex PL sentence, recursively

$$\begin{array}{ccc} P & Q & \neg (P \lor Q) \Leftrightarrow (\neg P \land \neg Q) \\ T & T & F & F \end{array}$$

T F F

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### Truth-Functional Analysis (03)

▶ Based on the basic truth tables, we can assign a semantic value to any complex PL sentence, recursively

P $\neg (P \lor Q) \Leftrightarrow (\neg P \land \neg Q)$ 

F

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## Truth-Functional Analysis (05)

▶ Based on the basic truth tables, we can assign a semantic value to any complex PL sentence, recursively

 $P \quad Q \qquad \neg (P \lor Q) \Leftrightarrow (\neg P \land \neg Q)$ 

F T F F

F F T

T

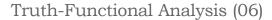
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T F F

T T T

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▶ Based on the basic truth tables, we can assign a semantic value to any complex PL sentence, recursively

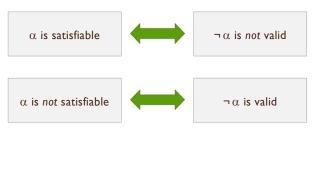
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# Valid and Satisfiable Sentences

▶ Key relationships to remember:



Valid and Satisfiable Sentences

- An expression like this is valid: true in any possible model that assigns truth values to propositional symbols
- In addition, a sentence like this is satisfiable, i.e. there exists some model that makes it true

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