

# Wumpus world characterization



- <u>Fully Observable</u> No only <u>local</u> perception
- <u>Deterministic</u>
- Episodic
- Static
- <u>Discrete</u>
- Single-agent

QU!

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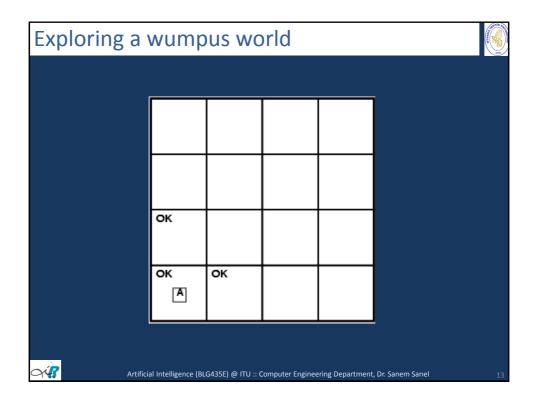
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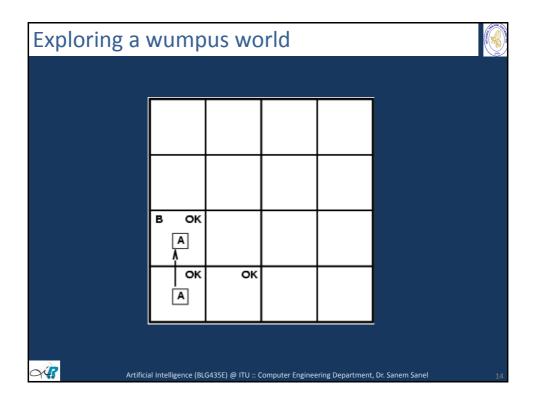


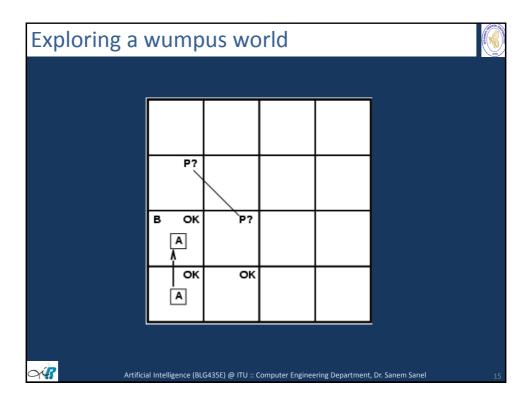
- Fully Observable No only local perception
- Deterministic Yes outcomes exactly specified
- <u>Episodic</u> No sequential at the level of actions
- Static Yes Wumpus and Pits do not move
- <u>Discrete</u> Yes
- <u>Single-agent</u> Yes Wumpus is essentially a natural feature

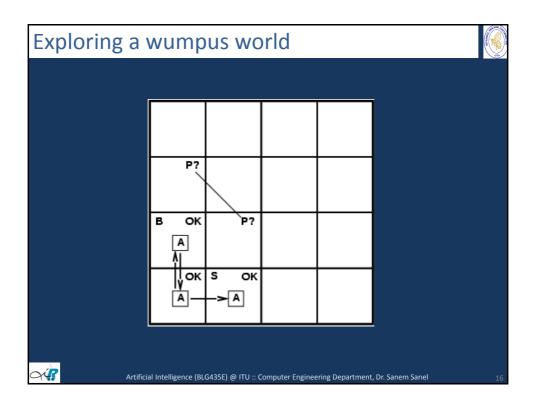
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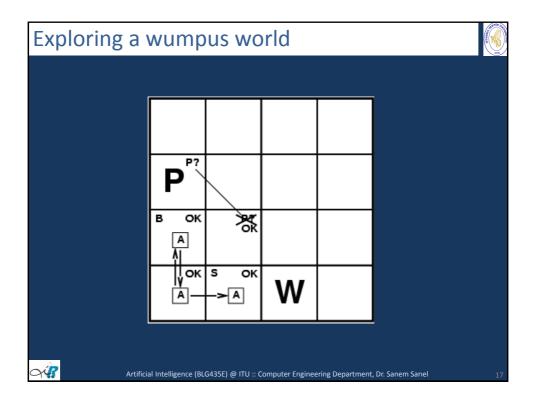
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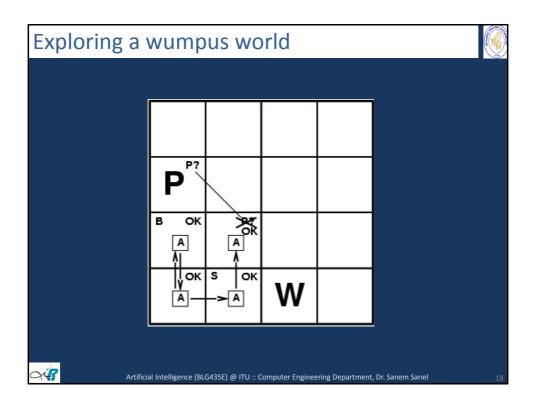


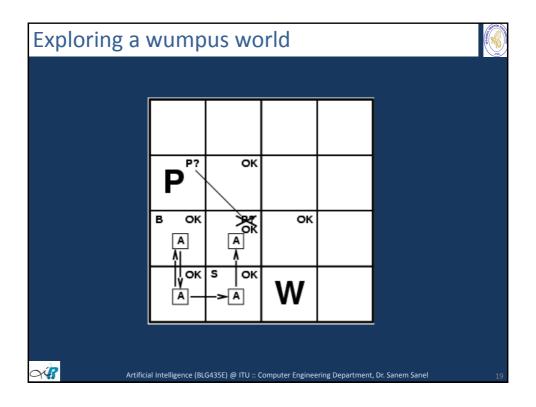


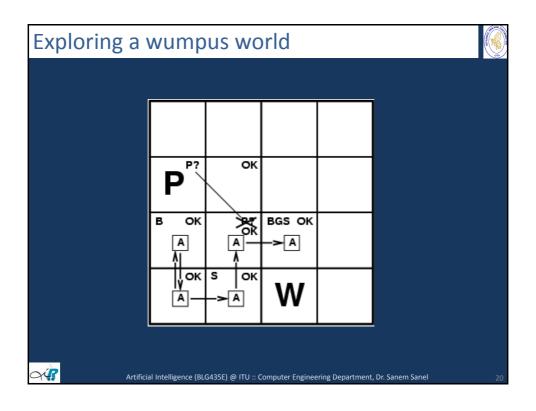












### Logic in general



- Logics are formal languages for representing information such that conclusions can be drawn
- Syntax defines the sentences in the language
- Semantics define the "meaning" of sentences;
  - i.e., define truth of a sentence in a world
- e.g., the language of arithmetic
  - $-x+2 \ge y$  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



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### **Entailment**



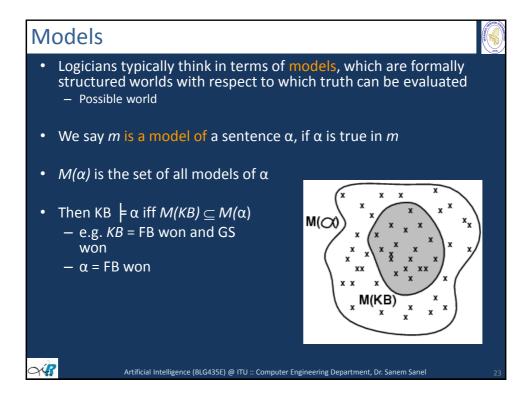
Entailment means that one thing follows from another:

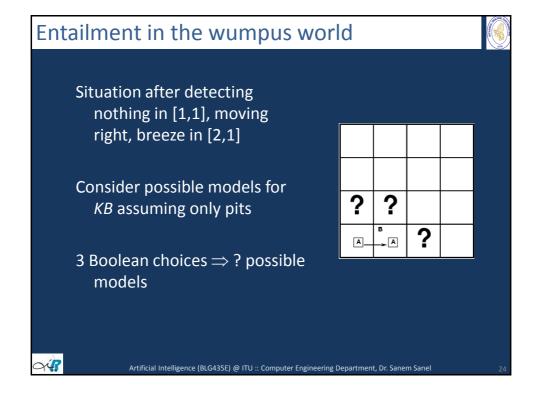
 $KB \models \alpha$ 

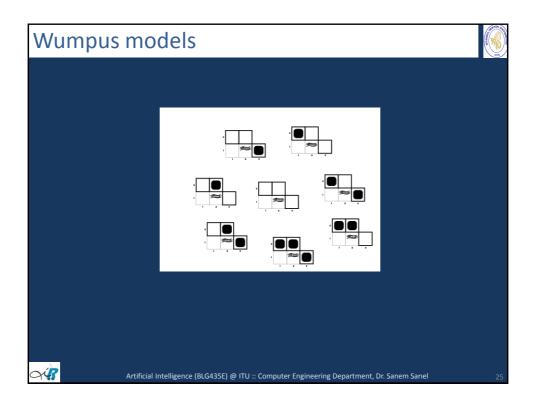
- Knowledge base KB entails sentence  $\alpha$  if and only if  $\alpha$  is true in all worlds where KB is true
  - e.g., the KB containing "FB won" and "GS won" entails "either GS won or FB won"
  - e.g., x+y = 4 entails 4 = x+y
  - Entailment is a relationship between sentences (i.e., syntax) that is based on semantics

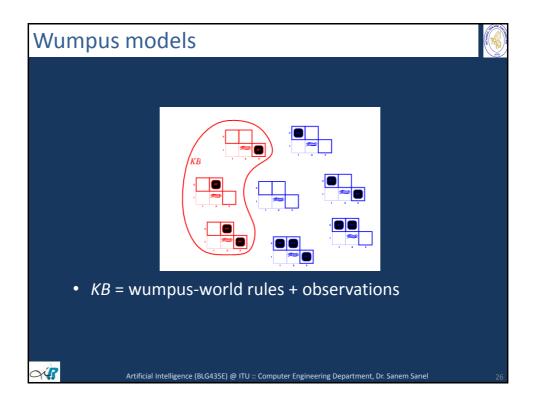
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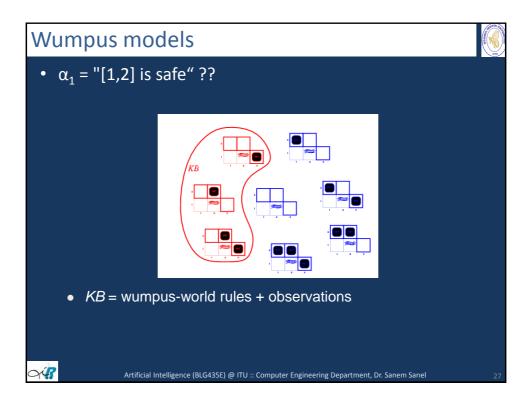
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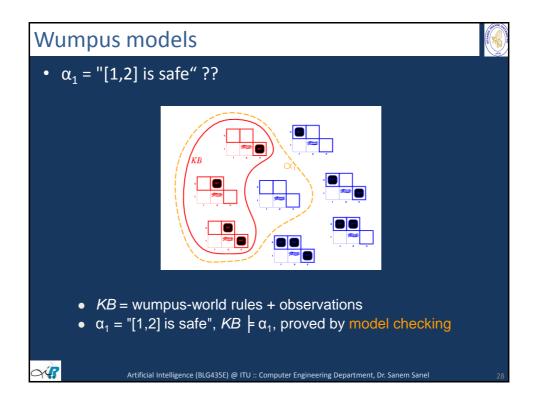


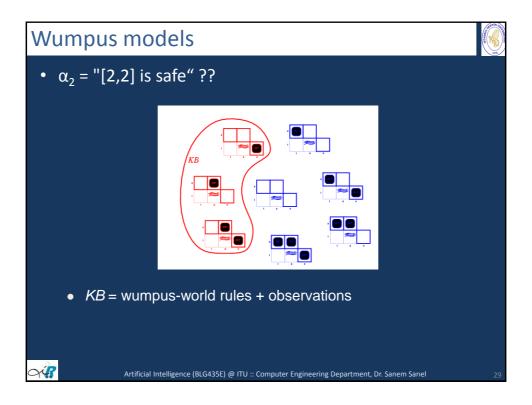


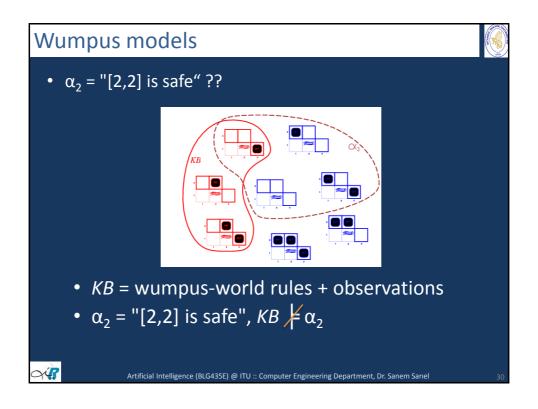












### Inference



- $KB \mid_{i} \alpha = \text{sentence } \alpha \text{ can be derived from } KB \text{ by an inference algorithm "i"}$ 
  - Needle in the haystack
- Soundness: "i" is sound if whenever  $KB \models_i \alpha$ , it is also true that  $KB \models \alpha$
- Completeness: "i" is complete if whenever  $KB \models \alpha$ , it is also true that  $KB \models_{i} \alpha$

OY!

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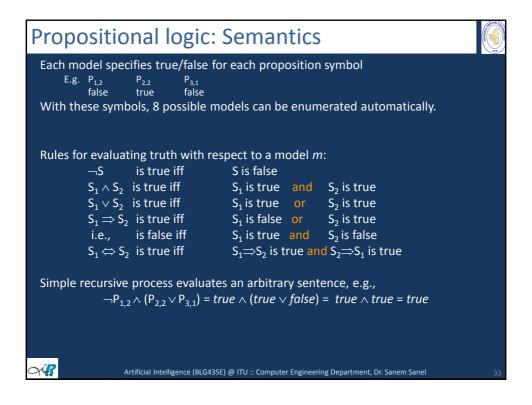
# Propositional logic: Syntax

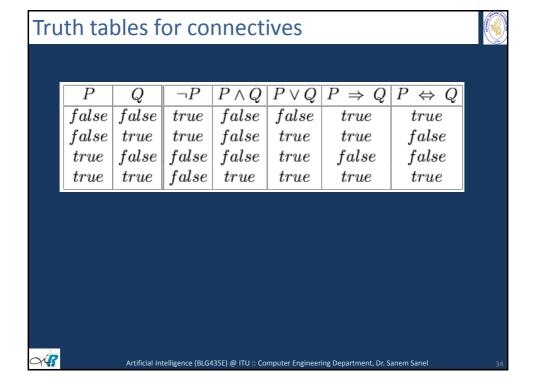


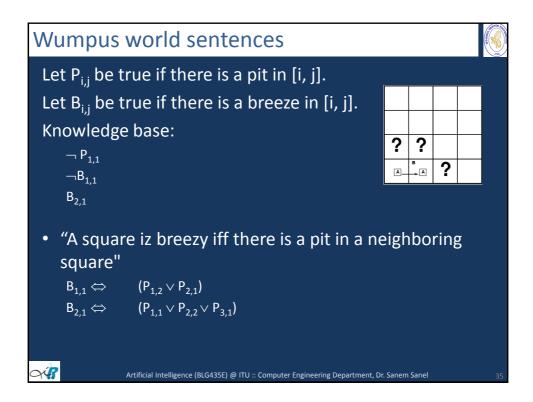
- Propositional logic is the simplest logic illustrates basic ideas
- The proposition symbols P<sub>1</sub>, P<sub>2</sub>, etc. are sentences
- Complex sentences are constructed from simpler sentences using parentheses and logical connectives
  - If S is a sentence, ¬S is a sentence (negation)
  - If  $S_1$  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)

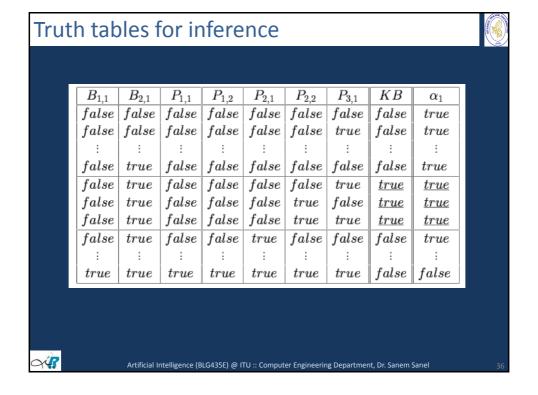
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### Inference by enumeration Depth-first enumeration of all models, sound and complete function TT-Entails? $(KB, \alpha)$ returns true or falseinputs: KB, the knowledge base, a sentence in propositional logic $\alpha$ , the query, a sentence in propositional logic $symbols \leftarrow$ a list of the proposition symbols in KB and $\alpha$ return TT-CHECK-ALL( $KB, \alpha, symbols, \{\}$ ) $\textbf{function} \ \mathsf{TT-Check-All}(KB, \alpha, symbols, model) \ \textbf{returns} \ true \ \mathsf{or} \ false$ if EMPTY?(symbols) then if PL-True?(KB, model) then return PL-True?( $\alpha$ , model) else return true // when KB is false, always return true $P \leftarrow FIRST(symbols)$ $rest \leftarrow Rest(symbols)$ **return** (TT-CHECK-ALL(KB, $\alpha$ , rest, $model \cup \{P = true\}$ ) and $\mathsf{TT\text{-}CHECK\text{-}ALL}(\mathit{KB},\alpha,\mathit{rest},\mathit{model}\ \cup \{P\ =\ \mathit{false}\ \}))$ For *n* symbols, time complexity is $O(2^n)$ , space complexity is O(n)Artificial Intelligence (BLG435E) @ ITU :: Computer Engineering Department, Dr. Sanem Sarıel

## **Propositional Theorem Proving**



- A sequence of applications of inference rules
- 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
    - heuristic search in model space (sound but incomplete)

e.g., min-conflicts-like hill-climbing algorithms



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### Logical equivalence



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

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(\alpha \wedge \beta) \equiv (\beta \wedge \alpha) \quad \text{commutativity of } \wedge \\ (\alpha \vee \beta) \equiv (\beta \vee \alpha) \quad \text{commutativity of } \vee \\ ((\alpha \wedge \beta) \wedge \gamma) \equiv (\alpha \wedge (\beta \wedge \gamma)) \quad \text{associativity of } \wedge \\ ((\alpha \vee \beta) \vee \gamma) \equiv (\alpha \vee (\beta \vee \gamma)) \quad \text{associativity of } \vee \\ \neg(\neg \alpha) \equiv \alpha \quad \text{double-negation elimination} \\ (\alpha \Rightarrow \beta) \equiv (\neg \beta \Rightarrow \neg \alpha) \quad \text{contraposition} \\ (\alpha \Rightarrow \beta) \equiv (\neg \alpha \vee \beta) \quad \text{implication elimination} \\ (\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)) \quad \text{biconditional elimination} \\ \neg(\alpha \wedge \beta) \equiv (\neg \alpha \vee \neg \beta) \quad \text{de Morgan} \\ \neg(\alpha \vee \beta) \equiv (\neg \alpha \wedge \neg \beta) \quad \text{de Morgan} \\ (\alpha \wedge (\beta \vee \gamma)) \equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) \quad \text{distributivity of } \wedge \text{ over } \vee \\ (\alpha \vee (\beta \wedge \gamma)) \equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) \quad \text{distributivity of } \vee \text{ over } \wedge \\ \text{Artificial Intelligence } (\text{BLG435E}) @ \text{ITU} :: Computer Engineering Department, Dr. Sanem Sariel}
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# 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.,  $A \land \neg A$ 

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

Reductio ad absurdum (reduction to an absurd thing) Proof by contradiction



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# Reasoning Patterns in Propositional Logic



Modus Ponens

$$- \underline{\alpha \Rightarrow \beta}, \quad \underline{\alpha}$$

And elimination

$$- \frac{\alpha \wedge \beta}{\alpha}$$

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### Resolution



An inference algorithm used with search algorithms

Conjunctive Normal Form (CNF)

conjunction of disjunctions of literals or clauses

e.g., 
$$(A \lor \neg B) \land (B \lor \neg C \lor \neg D)$$

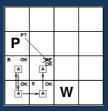
• Resolution inference rule (for CNF):

$$\frac{\mathit{l}_{1} \vee ... \vee \mathit{l}_{k}, \qquad m_{1} \vee ... \vee \mathit{m}_{n}}{\mathit{l}_{1} \vee ... \vee \mathit{l}_{l-1} \vee \mathit{l}_{l+1} \vee ... \vee \mathit{l}_{k} \vee m_{1} \vee ... \vee \mathit{m}_{l-1} \vee \mathit{m}_{l+1} \vee ... \vee \mathit{m}_{n}}$$

where  $l_i$  and  $m_i$  are complementary literals

e.g., 
$$P_{1,3} \lor P_{2,2}$$
,  $\neg P_{2,2}$ 

 Resolution is sound and complete for propositional logic



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### 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  $\neg$  inwards using de Morgan's rules and doublenegation:

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

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



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# Resolution algorithm



• Proof by contradiction, i.e., show  $KB \land \neg \alpha$  unsatisfiable

function PL-Resolution( $K\!B, \alpha$ ) returns true or false

inputs: KB, the knowledge base, a sentence in propositional logic  $\alpha$ , the query, a sentence in propositional logic

 $clauses \leftarrow$  the set of clauses in the CNF representation of  $KB \land \neg \alpha$   $new \leftarrow \{\ \}$ 

loop do

for each pair of clauses  $C_i$ ,  $C_j$  in clauses do

 $resolvents \leftarrow PL-Resolve(C_i, C_j)$ 

if resolvents contains the empty clause then return true

 $new \leftarrow new \cup resolvents$ 

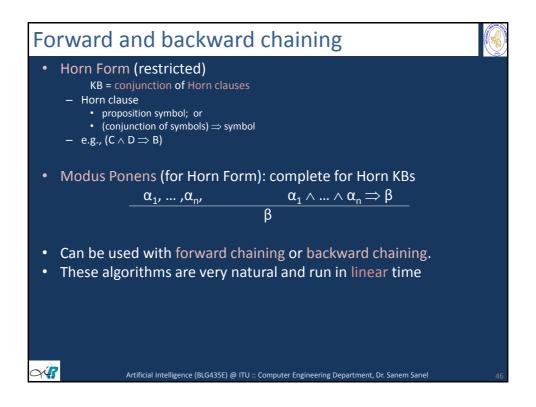
if  $new \subseteq clauses$  then return false

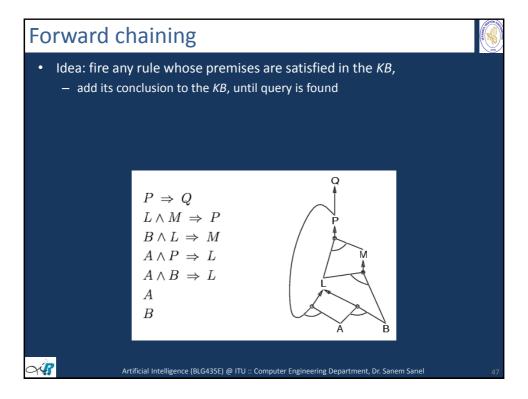
 $clauses \leftarrow clauses \cup new$ 

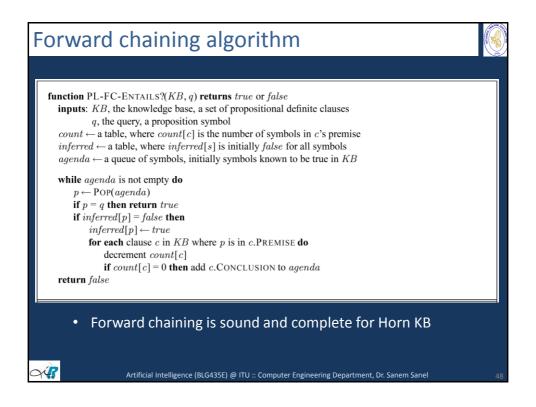


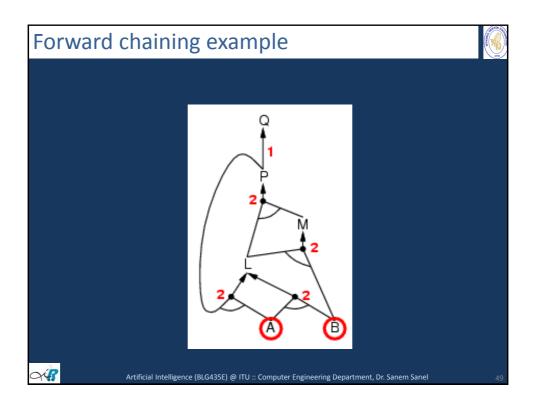
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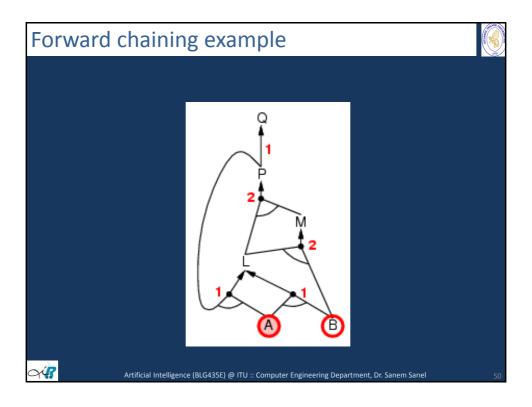
# Resolution example • $KB = (B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})) \wedge \neg B_{1,1} \quad \alpha = \neg P_{1,2}$ • $(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg P_{1,2} \vee B_{1,1}) \wedge (\neg P_{2,1} \vee B_{1,1}) \wedge \neg B_{1,1} \wedge P_{1,2}$ • $(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg P_{1,2} \vee P_{2,1}) \wedge (\neg P_{2,1} \vee B_{1,1}) \wedge \neg B_{1,1} \wedge P_{1,2}$ • $(\neg B_{1,1} \vee P_{1,2} \vee B_{1,1}) \wedge \neg B_{1,1} \vee P_{1,2} \vee P_{2,1} \wedge \neg P_{1,2} \vee P_{2,1} \wedge \neg P_{1,2} \wedge \neg P_{1,2} \wedge \neg P_{1,2} \wedge \neg P_{2,1} \wedge \neg P_{2$

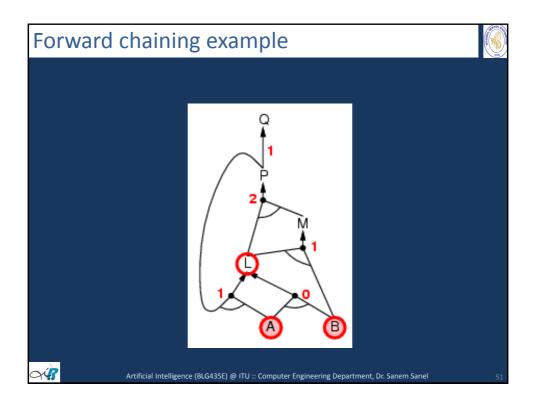


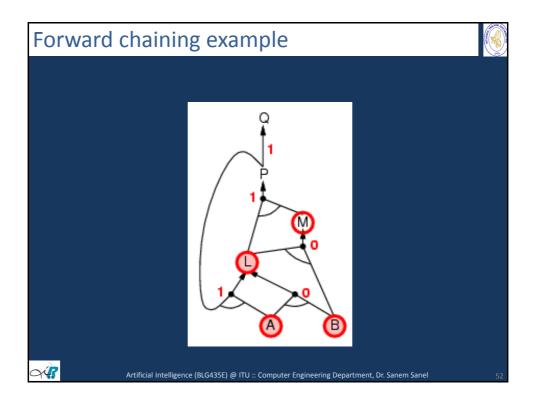


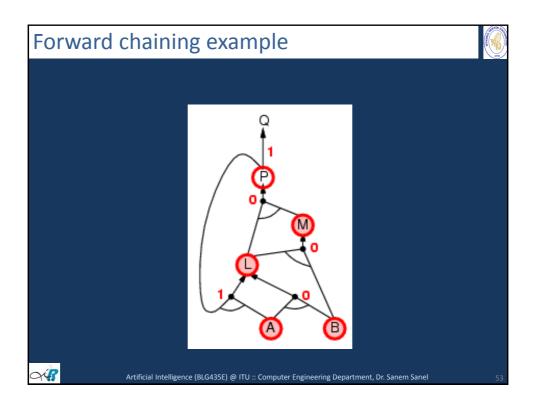


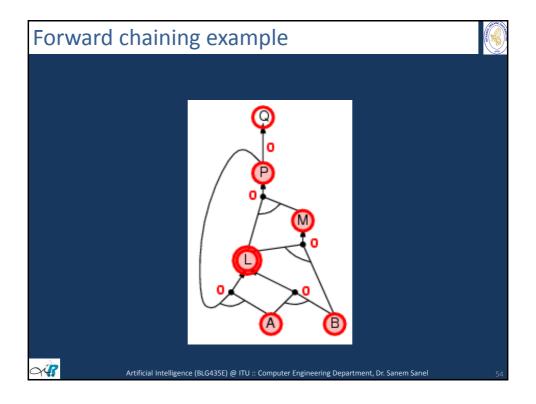


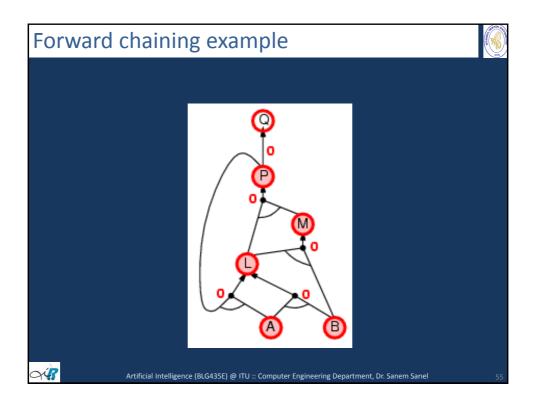


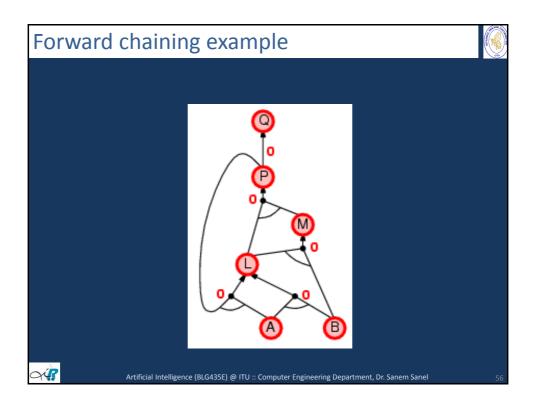


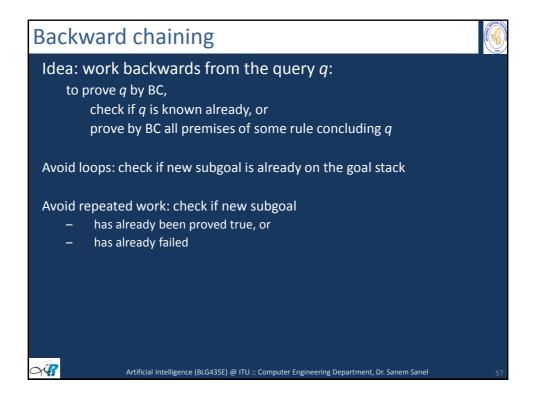


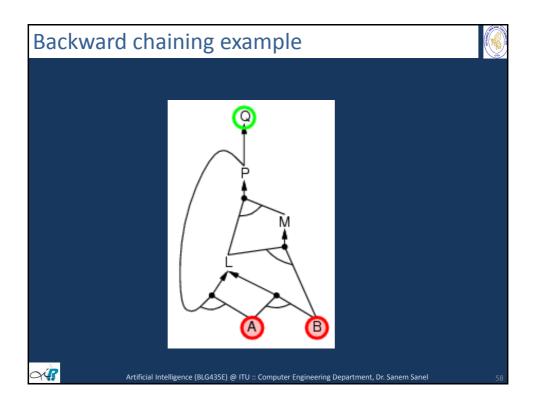


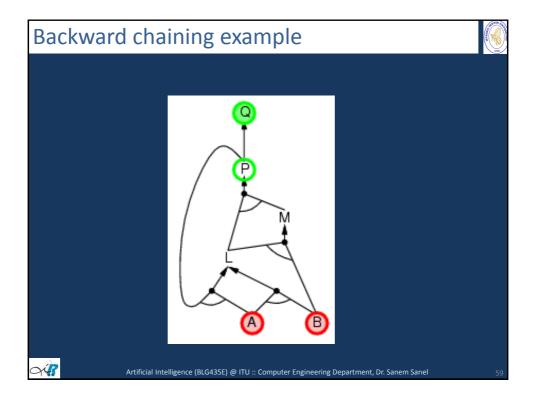


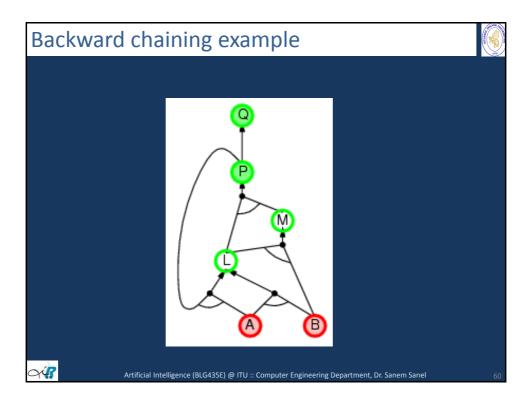


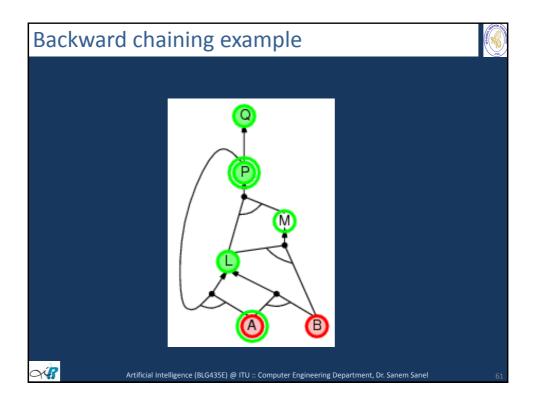


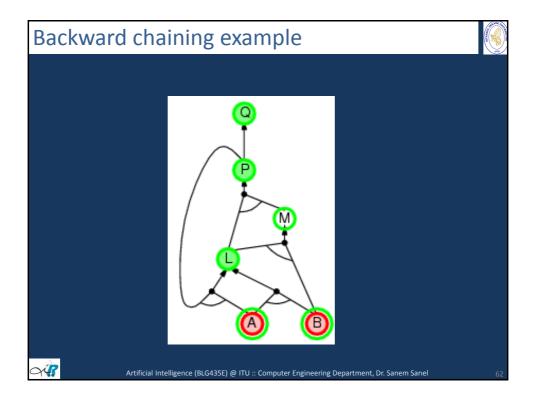


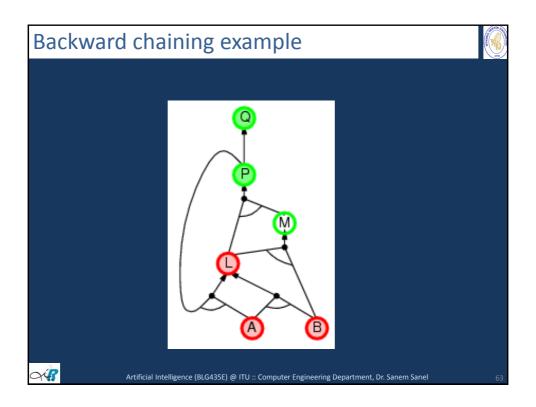


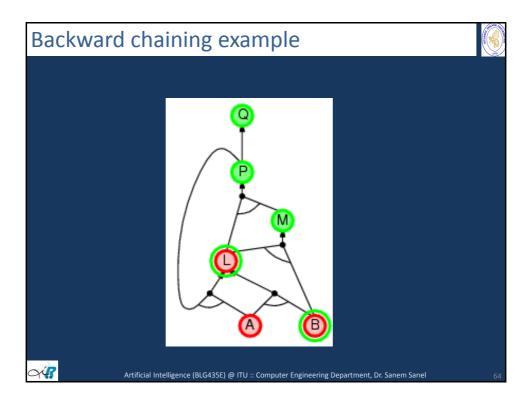


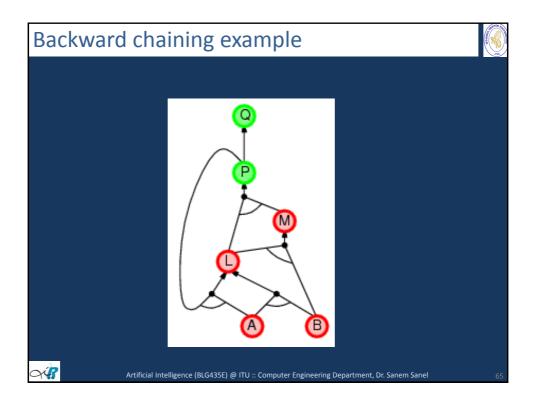


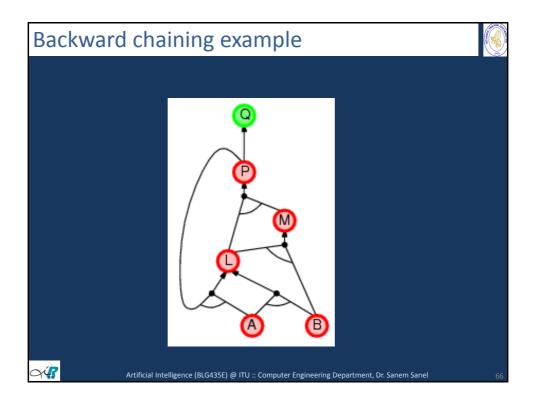


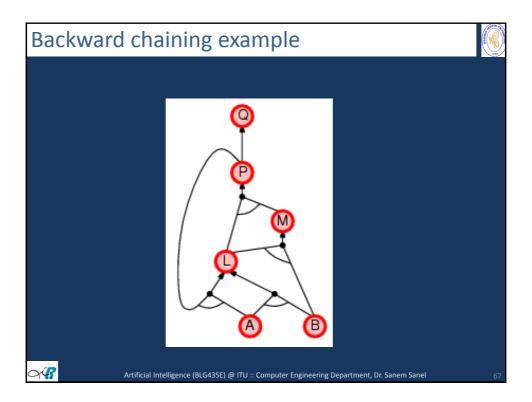












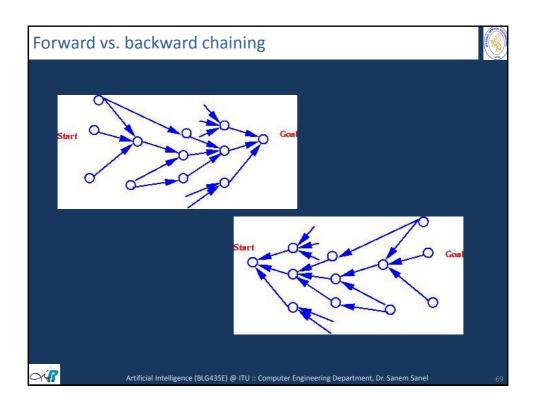
# Forward vs. backward chaining

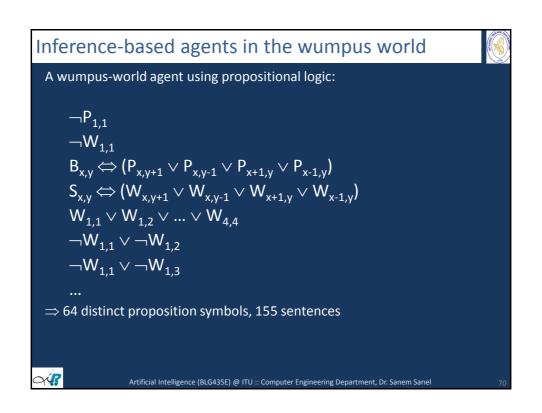


- FC is data-driven, automatic, unconscious processing
  - e.g., object recognition, routine decisions
  - May do lots of work that is irrelevant to the goal
- BC is goal-driven, appropriate for problem-solving
  - Where are my keys?
  - How can I get grade AA from AI?
- Complexity of BC can be much less than linear in size of KB

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### Expressiveness limitation of propositional logic

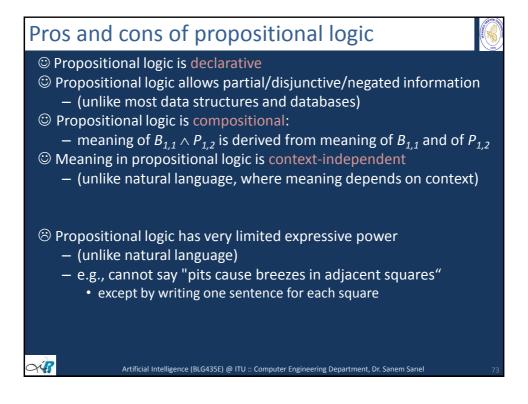


- KB contains "physics" sentences for every single square
- $L_{1.1} \wedge FacingRight \wedge Forward \Rightarrow L_{2.1}$
- For every time t and every location [x,y],
   L<sub>x,y</sub><sup>t</sup> ∧ FacingRight <sup>t</sup> ∧ Forward <sup>t</sup> ⇒ L<sub>x+1,y</sub> <sup>t+1</sup>
- Lacks the expressive power to deal with time, space and universal patterns of relationships among objects



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```
function Hybrid-Wumpus-Agent(percept) returns an action
        inputs: percept, a list, [stench,breeze,glitter,bump,scream]
        persistent: KB, a knowledge base, initially the atemporal "wumpus physics"
                                                     t, a counter, initially 0, indicating time
                                                     plan, an action sequence, initially empty
        {\tt Tell}(KB, {\tt Make-Percept-Sentence}(percept, t))
         TELL the KB the temporal "physics" sentences for time t
          safe \leftarrow \{[x, y] : Ask(KB, OK_{x,y}^t) = true\}
         \begin{split} & \text{if Ask}(KB, Glitter^t) = \textit{true then} \\ & \textit{plan} \leftarrow [\textit{Grab}] + \text{Plan-Route}(\textit{current}, \{[1,1]\}, \textit{safe}) + [\textit{Climb}] \end{split} 
        if plan is empty then
                  unvisited \leftarrow \{[x,y] : Ask(KB, L_{x,y}^{t'}) = false \text{ for all } t' \leq t\}
                   plan \leftarrow \texttt{PLAN-ROUTE}(current, unvisited \cap safe, safe)
        \begin{array}{l} \textbf{if } plan \text{ is empty and } \mathsf{ASK}(KB, HaveArrow^t) = true \textbf{ then} \\ possible\_wumpus \leftarrow \{[x,y]: \mathsf{ASK}(KB, \neg\ W_{x,y}) = false\} \end{array}
                   plan \leftarrow \texttt{PLAN-SHOT}(current, possible\_wumpus, safe)
        \label{eq:continuous} \begin{split} & \text{if } plan \text{ is empty then } \ /\! / \text{no choice but to take a risk} \\ & not\_unsafe \leftarrow \{[x,y]: \text{ASK}(KB, \neg OK_{x,y}^t) = false\} \\ & plan \leftarrow \text{PLAN-ROUTE}(current, unvisited \cap not\_unsafe, safe) \end{split}
        if plan is empty then % \frac{1}{2}\left( \frac{1}{2}\right) =\frac{1}{2}\left( \frac{1}{2}\right) +\frac{1}{2}\left( \frac{1}{2}\right) +\frac{
                  plan \leftarrow \texttt{PLAN-ROUTE}(current, \{[1, 1]\}, safe) + [Climb]
         action \leftarrow Pop(plan)
         Tell(KB, Make-Action-Sentence(action, t))
         t \leftarrow t + 1
        return action
function PLAN-ROUTE(current, goals, allowed) returns an action sequence
        inputs: current, the agent's current position
                                      goals, a set of squares; try to plan a route to one of them
                                         allowed, a set of squares that can form part of the route
         problem \leftarrow Route-Problem(current, goals, allowed)
         return A*-GRAPH-SEARCH(problem)
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```





# First-order logic



- Propositional logic assumes the world contains facts,
- First-order logic (like natural language) assumes the world contains
  - Objects: people, houses, numbers, colors, baseball games, wars, ...
  - Relations: red, round, prime, brother of, bigger than, part of, comes between, ...
  - Functions: father of, best friend, one more than, plus, ...
- "Squares neighboring the wumpus are smelly"

O'S

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# Syntax of FOL: Basic elements



- Constants: KingJohn, 2,...
- Predicates: Brother, >,...
- Functions: Sqrt, LeftLegOf,...
- Variables: x, y, a, b,...
- Connectives:  $\neg$ ,  $\Rightarrow$ ,  $\land$ ,  $\lor$ ,  $\Leftrightarrow$
- Equality: =
- Quantifiers: ∀,∃

Q'

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#### Atomic sentences



Term is a logical expression that refers to an object.

Term =  $function (term_1,...,term_n)$ or constant or variable

Atomic sentence =  $predicate (term_1,...,term_n)$ or  $term_1 = term_2$ <represent facts

• e.g., Married (Father(Richard), Mother(Richard))

OUR

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# **Complex sentences**



Complex sentences are made from atomic sentences using connectives

$$\neg S$$
,  $S_1 \land S_2$ ,  $S_1 \lor S_2$ ,  $S_1 \Rightarrow S_2$ ,  $S_1 \Leftrightarrow S_2$ ,

e.g.  $Sibling(KingJohn, Richard) \Rightarrow Sibling(Richard, KingJohn)$ 

QÚ

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## Truth in first-order logic



- Sentences are true with respect to a model and an interpretation
- Model contains objects (domain elements) and relations among them
- Interpretation specifies:

```
constant symbols → objects
predicate symbols → relations
function symbols → functional relations
```

An atomic sentence predicate(term<sub>1</sub>,...,term<sub>n</sub>) is true iff the objects referred to by term<sub>1</sub>,...,term<sub>n</sub> are in the relation referred to by predicate



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# Universal quantification



• ∀<*variables*> <*sentence*> (For all)

Everyone at ITU is smart:  $\forall x \text{ At}(x, \text{ITU}) \Rightarrow \text{Smart}(x)$ 

- $\forall x P$  is true in a model m iff P is true with x being each possible object in the model
  - Roughly speaking, equivalent to the conjunction of instantiations of P

```
At(KingJohn, ITU) \Rightarrow Smart(KingJohn) \land At(Richard, ITU) \Rightarrow Smart(Richard) \land At(ITU, ITU) \Rightarrow Smart(ITU) \land ...
```



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#### A common mistake to avoid



- Typically,  $\Rightarrow$  is the main connective with  $\forall$
- Common mistake: using ∧ as the main connective with ∀:

∀x At(x,ITU) ∧ Smart(x)
means "Everyone is at ITU and everyone is smart"



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# Existential quantification



- ∃<*variables*> <*sentence*> (There exists an x such that)
  We can name about some object without naming it
- Someone at ITU is smart:
- $\exists x \, At(x,ITU) \land Smart(x)$
- $\exists x P$  is true in a model m iff P is true with x being some possible object in the model
  - Roughly speaking, equivalent to the disjunction of instantiations of P

At(KingJohn,ITU) ∧ Smart(KingJohn) ∨ At(Richard,ITU) ∧ Smart(Richard) ∨ At(ITU,ITU) ∧ Smart(ITU) ∨ ...



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#### Another common mistake to avoid



- Typically, ∧ is the main connective with ∃
- Common mistake: using ⇒ as the main connective with ∃:

 $\exists x \; \mathsf{At}(\mathsf{x},\mathsf{ITU}) \Longrightarrow \mathsf{Smart}(\mathsf{x})$ 

is true if there is anyone who is not at ITU!

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# Properties of quantifiers



- $\forall x \forall y$  is the same as  $\forall y \forall x$
- $\exists x \exists y \text{ is the same as } \exists y \exists x$
- $\exists x \forall y$  is not the same as  $\forall y \exists x$
- ∃y ∀x Loves(x,y)
  - "There is someone who is loved by everyone"
- ∀x ∃y Loves(x,y)
  - "Everybody loves somebody"

Quantifier duality: each can be expressed using the other

 $\forall x \text{ Likes}(x,\text{IceCream})$   $\neg \exists x \neg \text{Likes}(x,\text{IceCream})$ 

∃x Likes(x,Broccoli)

 $\neg \forall x \neg Likes(x, Broccoli)$ 

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# **Equality**



- term<sub>1</sub> = term<sub>2</sub> is true under a given interpretation if and only if term<sub>1</sub> and term<sub>2</sub> refer to the same object
  - Father(John) = Henry
- e.g., definition of *Sibling* in terms of *Parent*:

```
\forall x,y \ Sibling(x,y) \Leftrightarrow [\neg(x = y) \land \exists m,f \neg (m = f) \land Parent(m,x) \land Parent(f,x) \land Parent(m,y) \land Parent(f,y)]
```

OY!

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# **Using FOL**



The kinship domain:

- Brothers are siblings
   ∀x,y Brother(x,y) ⇔ Sibling(x,y)
- One's mother is one's female parent
   ∀m,c Mother(c) = m ⇔ (Female(m) ∧ Parent(m,c))
- "Sibling" is symmetric
   ∀x,y Sibling(x,y) ⇔ Sibling(y,x)

QÚ

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### **Using FOL**



The set domain:

- $\forall s \text{ Set}(s) \Leftrightarrow (s = \{\}) \lor (\exists x, s_2 \text{ Set}(s_2) \land s = \{x \mid s_2\})$
- $\neg \exists x, s \{x \mid s\} = \{\}$
- $\forall x,s \ x \in s \Leftrightarrow s = \{x \mid s\}$
- $\forall x,s \ x \in s \Leftrightarrow [\exists y, s_2 \ (s = \{y \mid s_2\} \land (x = y \lor x \in s_2))]$
- $\forall s_1, s_2 \quad s_1 \subseteq s_2 \Leftrightarrow (\forall x \quad x \in s_1 \Rightarrow x \in s_2)$
- $\forall s_1, s_2 \ (s_1 = s_2) \Leftrightarrow (s_1 \subseteq s_2 \land s_2 \subseteq s_1)$
- $\forall x, s_1, s_2 \ x \in (s_1 \cap s_2) \Leftrightarrow (x \in s_1 \land x \in s_2)$
- $\forall x, s_1, s_2 \ x \in (s_1 \cup s_2) \Leftrightarrow (x \in s_1 \lor x \in s_2)$



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# Knowledge base for the wumpus world



• Suppose a wumpus-world agent is using a FOL KB and perceives a smell and a breeze (but no glitter) at *t=5*:

Tell(KB,Percept([Stench,Breeze,None],5))
Ask(KB,∃a BestAction(a,5))

- i.e., does the KB entail some best action at *t=5*?
- Answer: Yes, {a/Shoot} ← substitution (binding list)

QU!

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# Knowledge base for the wumpus world



- Given a sentence S and a substitution  $\sigma$ ,
- $\sigma$  denotes the result of plugging  $\sigma$  into S; e.g.,
  - S = Smarter(x,y)
  - $\sigma = \{x/Hillary, y/Bill\}$
  - $S\sigma = Smarter(Hillary,Bill)$
- Ask(KB,S) returns some/all  $\sigma$  such that KB  $\models \sigma$
- Perception
  - $\forall t,s,b \text{ Percept}([s,b,Glitter],t) \Rightarrow Glitter(t)$
- Reflex
  - $\forall$ t Glitter(t)  $\Rightarrow$  BestAction(Grab,t)



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# Deducing hidden properties



- ∀x,y,a,b Adjacent([x,y],[a,b]) ⇔
   [a,b] ∈ {[x+1,y], [x-1,y],[x,y+1],[x,y-1]}
- Home(Wumpus), unary predicate for Wumpus location

## Properties of squares:

•  $\forall$ s,t At(Agent,s,t)  $\land$  Breeze(t)  $\Rightarrow$  Breezy(s)



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## Deducing hidden properties



Squares are breezy near to a pit:

- Diagnostic rule
  - infer cause from effect

 $\forall$ s Breezy(s)  $\Rightarrow \exists$ r Adjacent(r,s)  $\land$  Pit(r)

- Causal rule
  - infer effect from cause

 $\forall$ r Pit(r)  $\Rightarrow$  [ $\forall$ s Adjacent(r,s)  $\Rightarrow$  Breezy(s)]

Systems that use casual rules are called model-based reasoning systems



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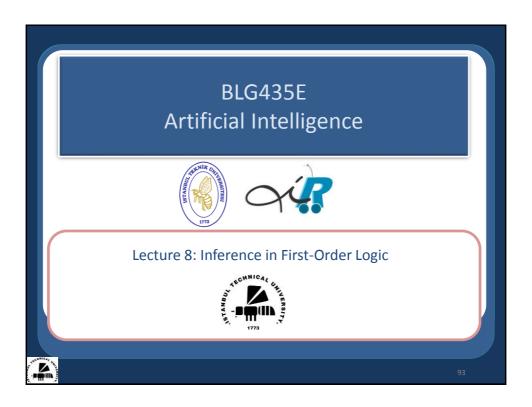
# Knowledge engineering in FOL

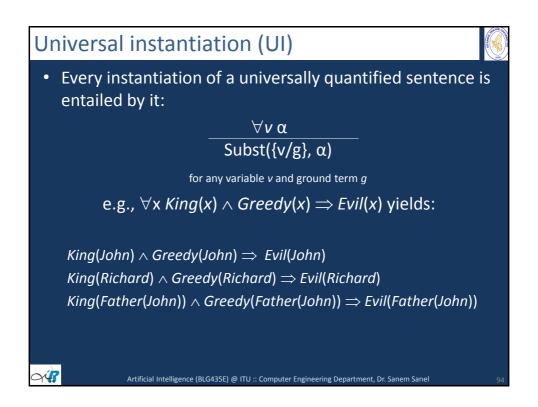


- 1. Identify the task
- 2. Assemble the relevant knowledge
- 3. Decide on a vocabulary of predicates, functions, and constants
- 4. Encode general knowledge about the domain
- 5. Encode a description of the specific problem instance
- 6. Pose queries to the inference procedure and get answers
- 7. Debug the knowledge base

OÚ?

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#### Existential instantiation (EI)



• For any sentence  $\alpha$ , variable  $\nu$ , and constant symbol k that does not appear elsewhere in the knowledge base:

 $\exists v \alpha$  Subst( $\{v/k\}, \alpha$ )

• e.g.,  $\exists x \ Crown(x) \land OnHead(x,John)$  yields:

 $Crown(C_1) \wedge OnHead(C_1, John)$ 

provided  $C_1$  is a new constant symbol, called a Skolem constant (Thoralf Skolem)



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# Reduction to propositional inference



Suppose the KB contains just the following:

 $\forall x \text{ King}(x) \land \text{Greedy}(x) \Rightarrow \text{Evil}(x)$ 

King(John)

Greedy(John)

Brother(Richard, John)

Instantiating the universal sentence in all possible ways, we have:

 $King(John) \wedge Greedy(John) \Rightarrow Evil(John)$ 

 $King(Richard) \land Greedy(Richard) \Rightarrow Evil(Richard)$ 

King(John)

Greedy(John)

Brother(Richard, John)

- The new KB is propositionalized: proposition symbols are:
  - King(John), Greedy(John), Evil(John), King(Richard), etc.

OU!

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#### Reduction contd.



- Every FOL KB can be propositionalized so as to preserve entailment
- A ground sentence is entailed by new KB iff entailed by the original KB
- Idea: propositionalize KB and query, apply resolution, return result
- Problem: with function symbols, there are infinitely many ground terms,
  - e.g., Father(Father(Father(John)))

OY!

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## Problems with propositionalization



- Propositionalization seems to generate lots of irrelevant sentences
- e.g., from:

 $\forall x \text{ King}(x) \land \text{Greedy}(x) \Rightarrow \text{Evil}(x)$ 

King(John)

∀y Greedy(y)

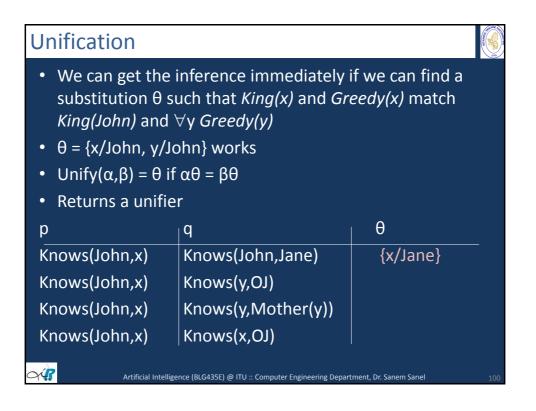
Brother(Richard, John)

 it seems obvious that Evil(John), but propositionalization produces lots of facts such as Greedy(Richard) that are irrelevant

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# Unification • We can get the inference immediately if we can find a substitution $\theta$ such that King(x) and Greedy(x) match King(John) and $\forall y Greedy(y)$ • $\theta = \{x/John, y/John\}$ works • Unify( $\alpha$ , $\beta$ ) = $\theta$ if $\alpha\theta$ = $\beta\theta$ θ p Knows(John, Jane) Knows(John,x) Knows(John,x) Knows(y,OJ) Knows(John,x) Knows(y,Mother(y)) Knows(x,OJ) Knows(John,x)



# Unification



- We can get the inference immediately if we can find a substitution  $\theta$  such that King(x) and Greedy(x) match King(John) and  $\forall y Greedy(y)$
- $\theta = \{x/John, y/John\}$  works
- Unify( $\alpha$ , $\beta$ ) =  $\theta$  if  $\alpha\theta$  =  $\beta\theta$

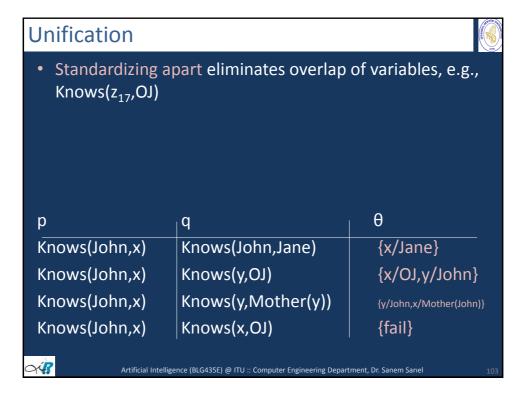
р	q	θ
Knows(John,x)	Knows(John,Jane)	{x/Jane}
Knows(John,x)	Knows(y,OJ)	{x/OJ,y/John}
Knows(John,x)	Knows(y,Mother(y))	
Knows(John,x)	Knows(x,OJ)	
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## Unification



- We can get the inference immediately if we can find a substitution θ such that King(x) and Greedy(x) match King(John) and ∀y Greedy(y)
- $\theta = \{x/John, y/John\}$  works
- Unify( $\alpha$ , $\beta$ ) =  $\theta$  if  $\alpha\theta$  =  $\beta\theta$

р	q	θ
Knows(John,x)	Knows(John,Jane)	{x/Jane}
Knows(John,x)	Knows(y,OJ)	{x/OJ,y/John}
Knows(John,x)	Knows(y,Mother(y))	{y/John,x/Mother(John)}
Knows(John,x)	Knows(x,OJ)	
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# Unification



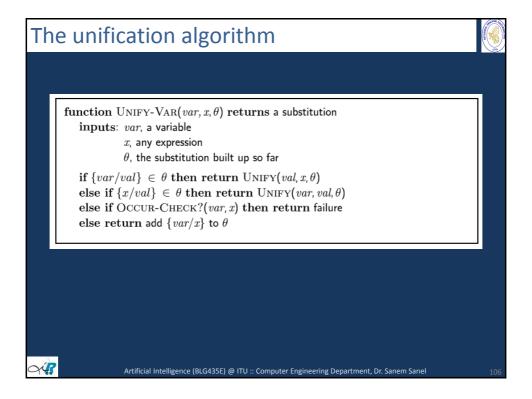
- To unify Knows(John,x) and Knows(y,z),
   θ = {y/John, x/z } or θ = {y/John, x/John, z/John}
- The first unifier is more general than the second.
- There is a single most general unifier (MGU) that is unique up to renaming of variables.
- MGU = { y/John, x/z }

QU!

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```
The unification algorithm

function UNIFY(x, y, θ) returns a substitution to make x and y identical inputs: x, a variable, constant, list, or compound y, a variable, constant, list, or compound θ, the substitution built up so far if θ = failure then return failure else if x = y then return θ else if VARIABLE?(x) then return UNIFY-VAR(x, y, θ) else if VARIABLE?(y) then return UNIFY-VAR(y, x, θ) else if COMPOUND?(x) and COMPOUND?(y) then return UNIFY(ARGS[x], ARGS[y], UNIFY(OP[x], OP[y], θ)) else if LIST?(x) and LIST?(y) then return UNIFY(REST[x], REST[y], UNIFY(FIRST[x], FIRST[y], θ)) else return failure
```



# Generalized Modus Ponens (GMP) $\frac{p_1', p_2', ..., p_n', (p_1 \land p_2 \land ... \land p_n \Rightarrow q)}{\text{SUBST}(\theta, q)}$ $\text{where } p_i'\theta = p_i \theta \text{ for all } i$ $p_1' \text{ is } King(John) \quad p_1 \text{ is } King(x)$ $p_2' \text{ is } Greedy(y) \quad p_2 \text{ is } Greedy(x)$ $\theta \text{ is } \{x/\text{John}, y/\text{John}\} \quad q \text{ is } Evil(x)$ SUBST(\theta, q) \text{ is } Evil(John) Artificial Intelligence (BLG435E) @ ITU :: Computer Engineering Department, Dr. Sanem Sanel

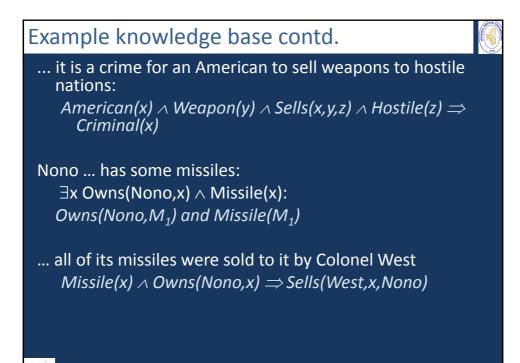
# Example knowledge base



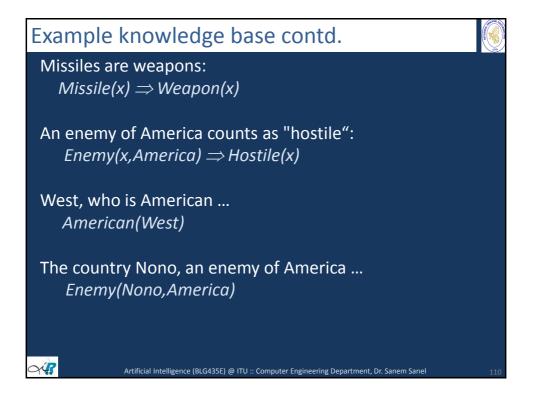
- The law says that it is a crime for an American to sell weapons to hostile nations. The country Nono, an enemy of America, has some missiles, and all of its missiles were sold to it by Colonel West, who is American.
- Prove that Colonel West is a criminal

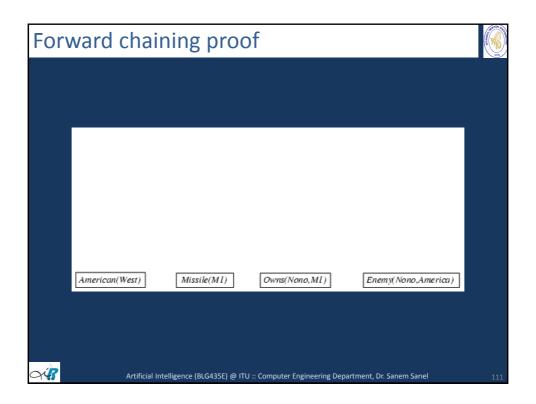
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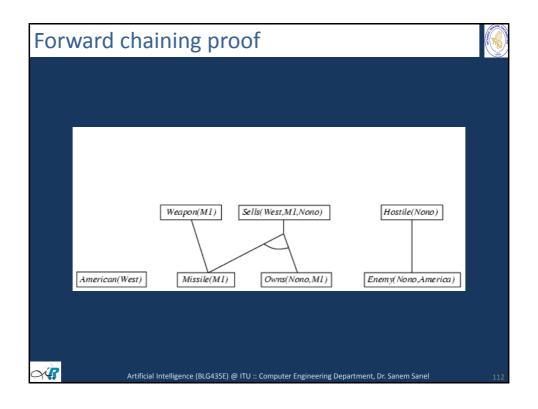
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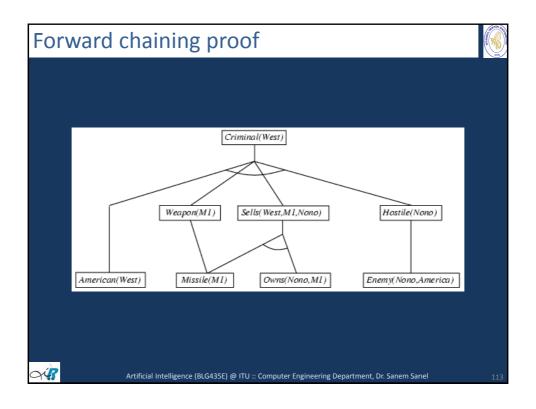


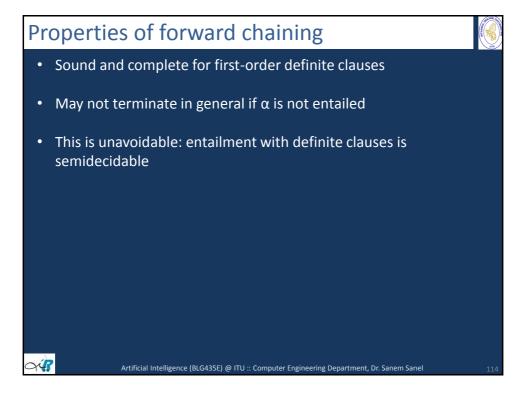
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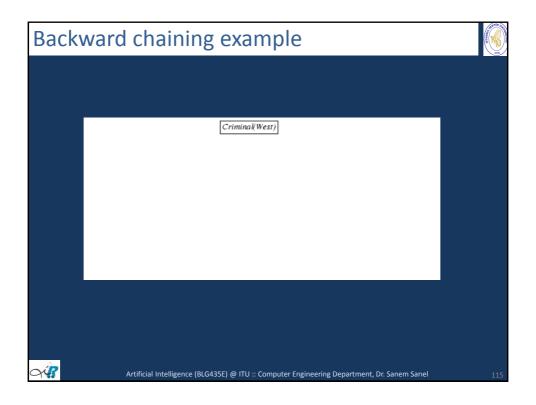


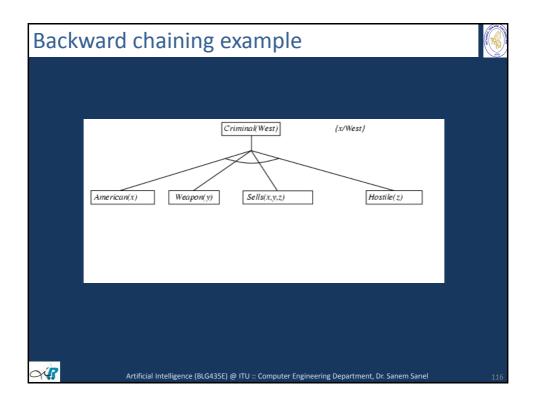


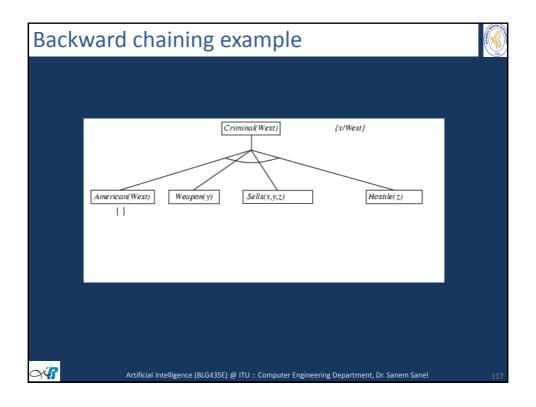


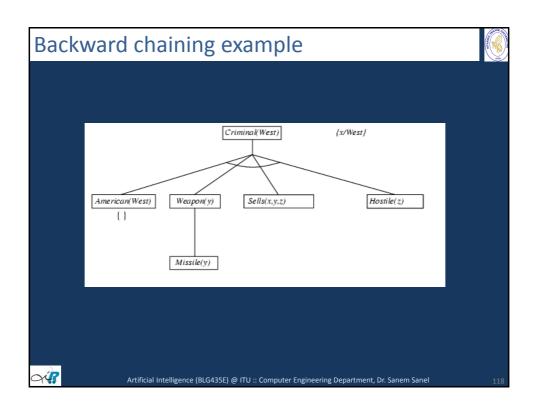


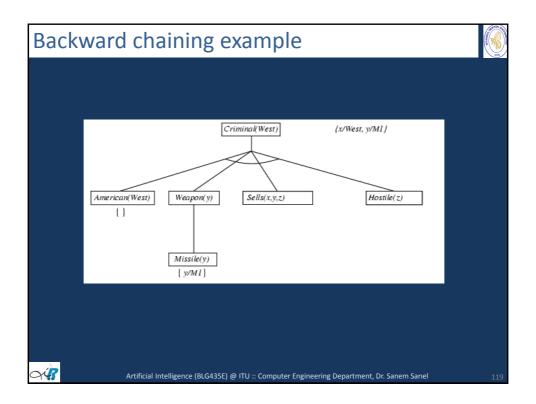


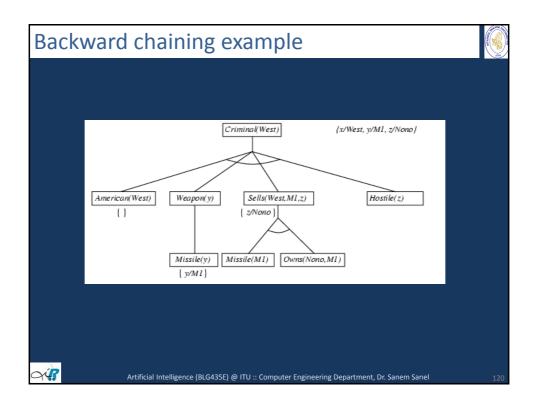


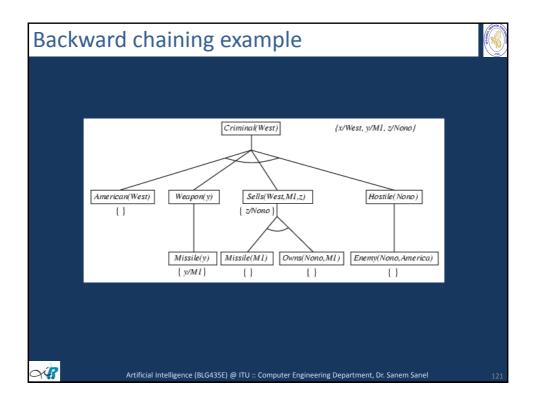


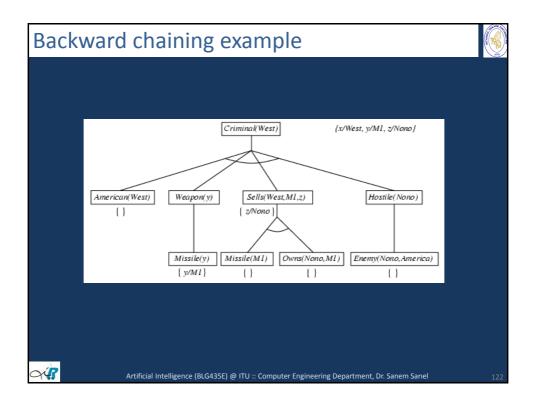












# Properties of backward chaining



- Depth-first recursive proof search: space is linear in size of proof
- Incomplete due to infinite loops
  - – ⇒ fix by checking current goal against every goal on stack
- Inefficient due to repeated subgoals (both success and failure)
  - $\Rightarrow$  fix using caching of previous results (extra space)
- · Widely used for logic programming



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# Resolution: brief summary



Full first-order version:

$$l_1 \vee \cdots \vee l_{\nu}$$

$$l_1 \vee \cdots \vee l_k$$
,  $m_1 \vee \cdots \vee m_n$ 

$$(\mathit{l}_{1} \vee \cdots \vee \mathit{l}_{i-1} \vee \mathit{l}_{i+1} \vee \cdots \vee \mathit{l}_{k} \vee \mathit{m}_{1} \vee \cdots \vee \mathit{m}_{j-1} \vee \mathit{m}_{j+1} \vee \cdots \vee \mathit{m}_{n})\theta$$

where Unify( $l_i$ ,  $\neg m_i$ ) =  $\theta$ .

 The two clauses are assumed to be standardized apart so that they share no variables:

$$\neg Rich(x) \lor Unhappy(x)$$

Rich(Ken)

Unhappy(Ken)

with  $\theta = \{x/Ken\}$ 

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#### Conversion to CNF



- Apply resolution steps to CNF(KB  $\wedge \neg \alpha$ ); complete for FOL
- Everyone who loves all animals is loved by someone:
   ∀x [∀y Animal(y) ⇒ Loves(x,y)] ⇒ [∃y Loves(y,x)]
- 1. Eliminate biconditionals and implications  $\forall x [\neg \forall y \neg Animal(y) \lor Loves(x,y)] \lor [\exists y Loves(y,x)]$
- 2. Move ¬ inwards:

```
\forall x [\exists y \neg (\neg Animal(y) \lor Loves(x,y))] \lor [\exists y Loves(y,x)] 
\forall x [\exists y \neg \neg Animal(y) \land \neg Loves(x,y)] \lor [\exists y Loves(y,x)] 
\forall x [\exists y Animal(y) \land \neg Loves(x,y)] \lor [\exists y Loves(y,x)]
```

OU!

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#### Conversion to CNF contd.



- 3. Standardize variables: each quantifier should use a different one  $\forall x \ [\exists y \ Animal(y) \land \neg Loves(x,y)] \lor [\exists z \ Loves(z,x)]$
- 4. Skolemize: a more general form of existential instantiation.
  Each existential variable is replaced by a Skolem function of the enclosing universally quantified variables:
- $\forall x [Animal(F(x)) \land \neg Loves(x,F(x))] \lor Loves(G(x),x)$
- 5. Drop universal quantifiers:  $[Animal(F(x)) \land \neg Loves(x,F(x))] \lor Loves(G(x),x)$
- 6. Distribute  $\vee$  over  $\wedge$  : [ $Animal(F(x)) \vee Loves(G(x),x)$ ]  $\wedge$  [ $\neg Loves(x,F(x)) \vee Loves(G(x),x)$ ]

OY!

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# Example knowledge base contd.



... it is a crime for an American to sell weapons to hostile nations:

American(x)  $\land$  Weapon(y)  $\land$  Sells(x,y,z)  $\land$  Hostile(z)  $\Rightarrow$  Criminal(x)

Nono ... has some missiles:

 $\exists x \ Owns(Nono,x) \land Missile(x):$   $Owns(Nono,M_1) \ and \ Missile(M_1)$ 

... all of its missiles were sold to it by Colonel West  $Missile(x) \land Owns(Nono,x) \Rightarrow Sells(West,x,Nono)$ 



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# Example knowledge base contd.



Missiles are weapons:

 $Missile(x) \Rightarrow Weapon(x)$ 

An enemy of America counts as "hostile":  $Enemy(x,America) \Rightarrow Hostile(x)$ 

West, who is American ...

American(West)

The country Nono, an enemy of America ... Enemy(Nono,America)



Artificial Intelligence (BLG435E) @ ITU :: Computer Engineering Department, Dr. Sanem Sariel

