

Bookkeeping

- Notes
 - HW4 out tonight—this will be the last homework (because projects)
 - NO CLASS NEXT THURSDAY—please work on projects!
- Today's class
 - A little about higher-order logics
 - More on knowledge-based agents
 - Situations reasoning over time
 - Inference in knowledge bases, 5 ways
- And a PSA (next)

Early Voting: Ends Thursday 10/31



https://elections.maryland.gov/voting /early_voting.html

Early Voting

2024 Presidential General Election

You can vote in person during early voting. Early voting will be held from Thursday, October 24, 2024 through Thursday, October 31, 2024. Early voting centers will be open from 7 am to 8 pm.

Who can vote early?

Any person that is registered to vote can vote during early voting. Any person that is eligible to register to vote can vote during early voting.

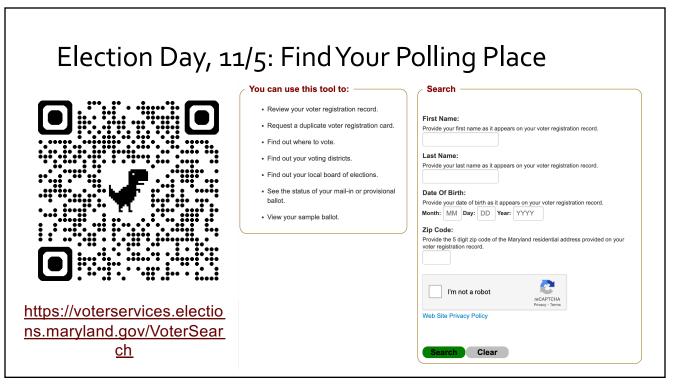
Registered voters have always been able to vote during early voting, but now individuals who are eligible but not yet registered can register and vote.

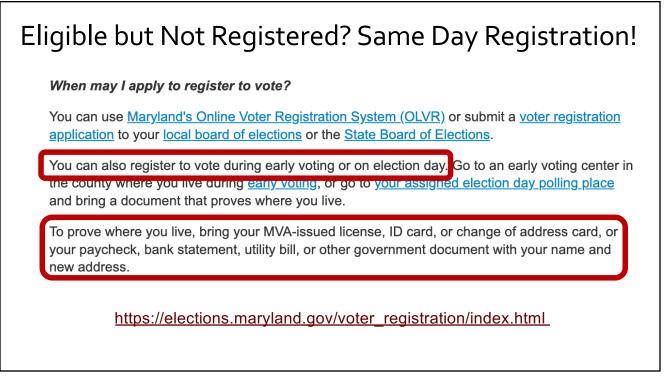
To register and vote during early voting, go to an early voting center in the county where you live and bring a document that proves where you live. This document can be your MVA-issued license, ID card, change of address card, your paycheck, bank statement, utility bill, or other government document with your name and new address. You will be able to register to vote and vote.

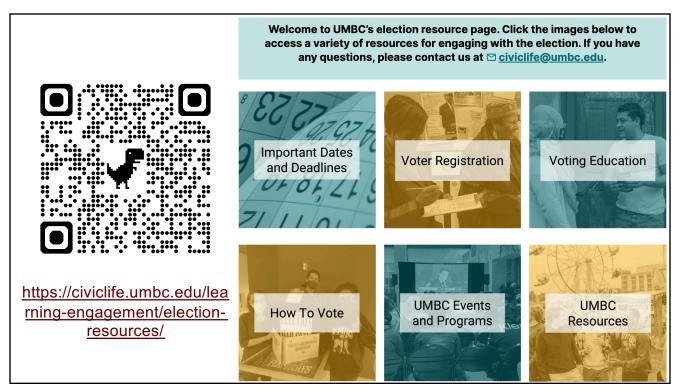
When can I vote early?

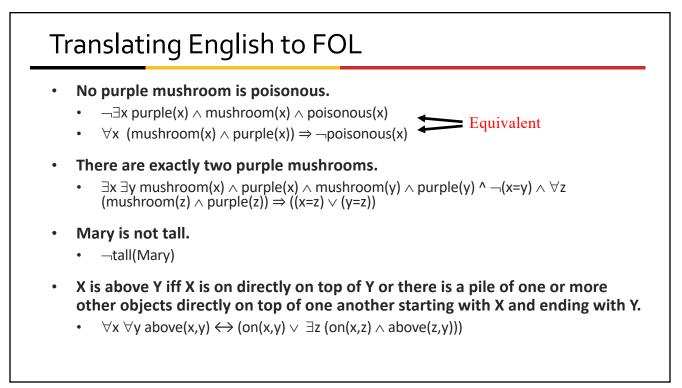
For the 2024 Presidential General Election, early voting will be available from Thursday, October 24, 2024 through Thursday, October31, 2024 (including Saturday and Sunday) from 7 am to 8 pm.













Semantics of FOL

- Domain M: the set of all objects in the world (of interest)
- Interpretation I: includes
 - Assign each constant to an object in M
 - Define each function of n arguments as a mapping Mⁿ => M
 - Define each predicate of n arguments as a mapping Mⁿ => {T, F}
 - Therefore, every ground predicate with any instantiation will have a truth value
 - In general there is an infinite number of interpretations because |M| is infinite
- Define logical connectives: ~, ^, v, =>, <=> as in PL
- Define semantics of (∀x) and (∃x)
 - $(\forall x) P(x)$ is true iff P(x) is true under all interpretations
 - $(\exists x) P(x)$ is true iff P(x) is true under some interpretation

Terminology

- **Model**: an interpretation of a set of sentences such that every sentence is True
- A sentence is
 - Satisfiable if it is true under some interpretation
 - Valid if it is true under all possible interpretations
 - **Inconsistent** if there does not exist any interpretation under which the sentence is true
- Logical consequence: S ⊨ X if all models of S are also models of X

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Axioms, Definitions and Theorems

- Axioms are facts and rules that attempt to capture all of the (important) facts and concepts about a domain; axioms can be used to prove **theorems**
 - Mathematicians don't want any unnecessary (dependent) axioms –ones that can be derived from other axioms
 - Dependent axioms can make reasoning faster, however
 - Choosing a good set of axioms for a domain is a kind of design problem
- A definition of a predicate is of the form "p(X) ↔ …" and can be decomposed into two parts
 - Necessary description: " $p(x) \Rightarrow ...$ "
 - Sufficient description "... $\Rightarrow p(x)$ "
 - Some concepts don't have complete definitions (e.g., person(x))

Necessary and Sufficient

- p is necessary for q
 - $\neg p \Rightarrow \neg q$ ("no p, no q!")
- p is **sufficient** for q
 - $p \Rightarrow q$ ("p is all we need to know!")
- Note that $\neg p \Rightarrow \neg q$ is equivalent to $q \Rightarrow p$
- So if p is necessary and sufficient for q, then p iff q.



More on Definitions

- Examples: define father(x, y) by parent(x, y) and male(x)
 - parent(x, y) is a **necessary (but not sufficient**) description of father(x, y)
 - father(x, y) \Rightarrow parent(x, y)
 - parent(x, y) ^ male(x) ^ age(x, 35) is a sufficient (but not necessary) description of father(x, y):
 - father(x, y) \leftarrow parent(x, y) ^ male(x) ^ age(x, 35)
 - parent(x, y) ^ male(x) is a **necessary and sufficient** description of father(x, y)
 - parent(x, y) ^ male(x) \leftrightarrow father(x, y)

Converting FOL to CNF

- Eliminate biconditionals and implications
- Move ¬ inwards
- Standardize variables apart by renaming them: each quantifier should use a different variable
- **Skolemize:** each existential variable is replaced by a Skolem constant or Skolem function of the enclosing universally quantified variables.
 - For instance, $\exists x \operatorname{Rich}(x)$ becomes $\operatorname{Rich}(G1)$ where G1 is a new Skolem constant
 - "Everyone has a heart" [∀x Person(x) ⇒ ∃y Heart(y)∧Has(x,y)] becomes
 ∀x Person(x) ⇒ Heart(H(x)) ∧ Has(x, H(x)), where H is a new symbol (Skolem function)
- Drop universal quantifiers
 - For instance, ∀ x Person(x) becomes Person(x).
- Distribute ∧ over ∨

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Summary: First Order Logic (FOL)

- Uses the same logical symbols as Propositional Logic (PL)
- Adds: variables, quantification, predicates and functions
 - Names of terms: constants, variables, predicates, functions
- Existential and universal quantifiers can be used to create rules
- Need to be able to translate English to and from FOL
- Some extensions...

Higher-Order Logic

- FOL only allows to quantify over variables, and variables can only range over objects.
- HOL allows us to quantify over relations
- Example: (quantify over functions)
 - "two functions are equal iff they produce the same value for all arguments"
 - $\forall f \forall g (f = g) \leftrightarrow (\forall x f(x) = g(x))$
- Example: (quantify over predicates)
 - $\forall r \text{ transitive}(r) \Rightarrow (\forall xyz) r(x,y) \land r(y,z) \Rightarrow r(x,z))$
- More expressive, but undecidable.

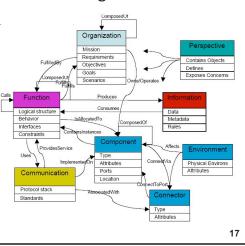


Expressing Uniqueness

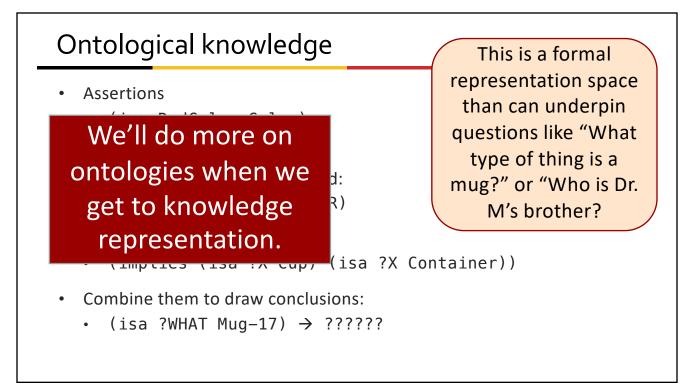
- Sometimes we want to say that there is a single, unique object that satisfies a certain condition
- "There exists a unique x such that king(x) is true"
 - $\exists x \text{ king}(x) \land \forall y \text{ (king}(y) \Rightarrow x=y)$
 - $\exists x \text{ king}(x) \land \neg \exists y \text{ (king}(y) \land x \neq y)$
- Iota operator: " $\iota x P(x)$ " means "the unique x such that p(x) is true"
 - "The unique ruler of Freedonia is dead"
 - dead(i x ruler(freedonia,x))

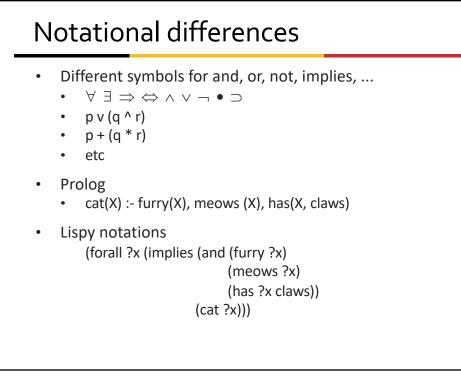
Knowledge bases/ontologies

- Ontology: the study of what there is—an inventory of what exists
- An ontology: a hierarchical categorization system for things in the world
- A formally represented corpus of knowledge
 - Defined by some grammar
 - Incorporates *rules* (implicitly or explicitly)
 - Not divided into tables: more like a graph
 - Often hierarchical
- Usually incorporate background knowledge (not purely domain-specific)
 - Although many are in a domain, such as biology









 Everything is bitter or sweet. Either everything is bitter or everything is sweet. There is somebody who is loved by everyone. Nobody is loved by no one. If someone is noisy, everybody 	 8. No frog is green. 9. Some frogs are not green. 10. A mechanic likes Bob. 11. A mechanic likes herself. 12. Every mechanic likes Bob. 13. Some mechanic likes every nurse.
is annoyed 6. Frogs are green. 7. Frogs are not green.	14. There is a mechanic who is liked by every nurse.

Exercise: FOL translation

- 1. $\forall x \text{ (bitter(x)} \lor \text{sweet(x))}$
- 2. $\forall x \text{ (bitter(x))} \lor \forall x \text{ (sweet(x))}$
- 3. $\exists x \forall y (loves(y,x))$
- 4. $\neg \exists x \neg \exists y (loves(y,x))$
- 5. $\exists x (noisy(x)) \Rightarrow \forall y(annoyed(y))$
- 6. $\forall x \text{ (frog}(x) \Rightarrow \text{green}(x))$
- 7. $\forall x \text{ (frog}(x) \Rightarrow \neg \text{green}(x))$

- 8. $\neg \exists x (frog(x) \land green(x))$
- 9. $\exists x (frog(x) \land \neg green(x))$
- 10. $\exists x (mech.(x) \land likes(x, Bob))$
- 11. $\exists x (mech.(x) \land likes(x, x))$
- 12. $\forall x \text{ (mech.}(x) \Rightarrow \text{likes}(x, \text{Bob}))$
- 13. $\exists x \forall y (mech(x) \land nurse(y)$ ⇒ likes(x, y))
- 14. $\exists x (mech(x) \land \forall y (nurse(y)))$ $\Rightarrow likes(y, x))$

Exercises: disi.unitn.it/~bernardi/Courses/LSNL/Slides/fl1.pdf

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A Note on Common Sense Reasoning – example adapted from Lenat

- You are told: John drove to the grocery store and bought a pound of noodles, a pound of ground beef, and two pounds of tomatoes.
 - Is John 3 years old?
 - Is John a child?
 - What will John do with the purchases?
 - Did John have any money?
 - Does John have less money after going to the store?
 - Did John buy at least two tomatoes?
 - Were the tomatoes made in the supermarket?
 - Did John buy any meat?
 - Is John a vegetarian?
 - Will the tomatoes fit in John's car?
- Can Propositional Logic support these inferences?

A Note on Common Sense Reasoning

- There are a number of inferences and conclusions that we can draw that depend on *background knowledge*
- We refer to this background knowledge as "common sense"
- Can be represented as a set of statements in a knowledge base
 - Only adults can drive
 - Tomatoes weigh less than 2 pounds
 - Purchasing involves spending money
 - When you spend money, you no longer have it [etc.]
- Given these statements, we can **infer** the answers to the questions

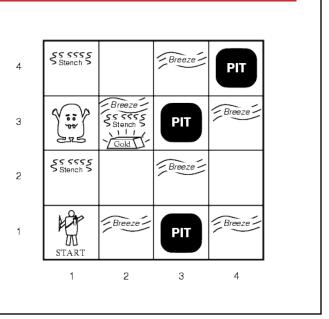


Logical Agents for Wumpus World

- Three (non-exclusive) agent architectures:
 - Reflex agents
 - Have rules that classify situations, specifying how to react to each possible situation
 - Model-based agents
 - Construct an internal model of their world
 - Goal-based agents
 - Form goals and try to achieve them

A Typical Wumpus World

- The agent always starts in the field [1,1].
- The task of the agent is to find the gold, return to the field [1,1] and climb out of the cave.



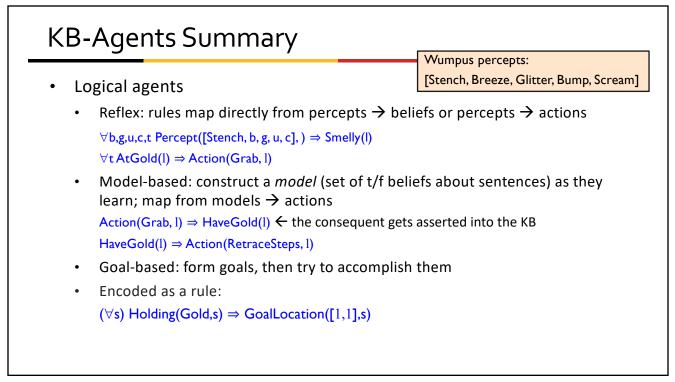
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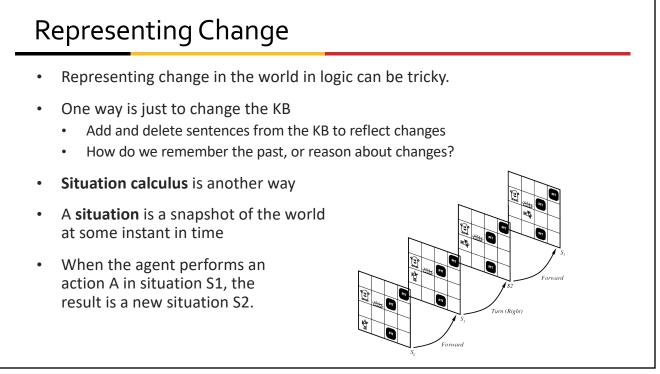
A Simple Reflex Agent

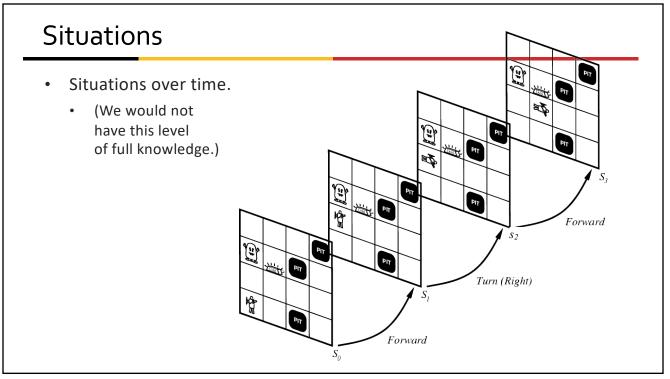
- Rules to map percepts into observations:
 - \forall b,g,u,c,l Percept([Stench, b, g, u, c], l) \Rightarrow Smelly(l)
 - \forall s,g,u,c,I Percept([s, Breeze, g, u, c], I) \Rightarrow Breezy(I)
 - \forall s,b,u,c,l Percept([s, b, Glitter, u, c], l) \Rightarrow AtGold(l)
- Rules to select an action given observations:
 - $\forall I AtGold(I) \Rightarrow Action(Grab, I)$

A Simple Reflex Agent

- Some difficulties:
- Climb?
 - There is no percept that indicates the agent should climb out position and holding gold are not part of the percept sequence
- Loops?
 - The percept will be repeated when you return to a square, which should cause the same response (unless we maintain some **internal model of the world**)







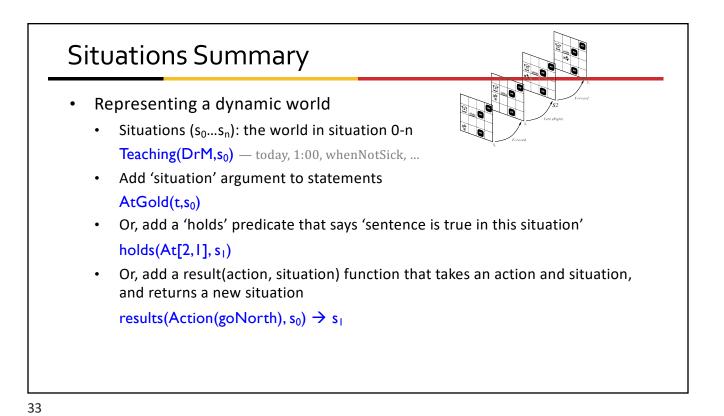
Situation Calculus

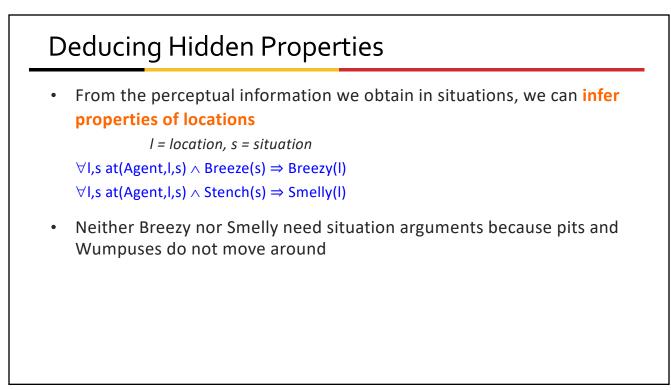
- A situation is:
 - A snapshot of the world
 - At an interval of time
 - During which nothing changes
- Every true or false statement is made wrt. a situation
 - Add situation variables to every predicate.
 - at(Agent,1,1) becomes at(Agent,1,1,s0): at(Agent,1,1) is true in situation (i.e., state) s0.

Situation Calculus

- Alternatively, add a special 2nd-order predicate, holds(f,s), that means "f is true in situation s." E.g., holds(at(Agent,1,1),s0)
- Or: add a new function, result(a,s), that maps a situation s into a new situation as a result of performing action a. For example, result(forward, s) is a function that returns the successor state (situation) to s
- Example: The action *agent-walks-to-location-y* could be represented by

 $(\forall x)(\forall y)(\forall s) (at(Agent,x,s) \land \neg onbox(s)) \Rightarrow at(Agent,y,result(walk(y),s))$





Deducing Hidden Properties II

- We need to write some rules that relate various aspects of a single world state (as opposed to across states)
- There are two main kinds of such rules:

Deducing Hidden Properties II

- We need to write some rules that relate various aspects of a single world state (as opposed to across states)
- There are two main kinds of such rules:
 - Causal rules reflect assumed direction of causality: (∀I1,I2,s) At(Wumpus,I1,s) ∧ Adjacent(I1,I2) ⇒ Smelly(I2) (∀ I1,I2,s) At(Pit,I1,s) ∧ Adjacent(I1,I2) ⇒ Breezy(I2)
- Systems that reason with causal rules are called model-based reasoning systems

Deducing Hidden Properties II

- We need to write some rules that relate various aspects of a single world state (as opposed to across states)
- There are two main kinds of such rules:

Diagnostic rules infer the presence of hidden properties directly from the percept-derived information. We have already seen two:
 (∀ I,s) At(Agent,I,s) ∧ Breeze(s) ⇒ Breezy(I)
 (∀ I,s) At(Agent,I,s) ∧ Stench(s) ⇒ Smelly(I)

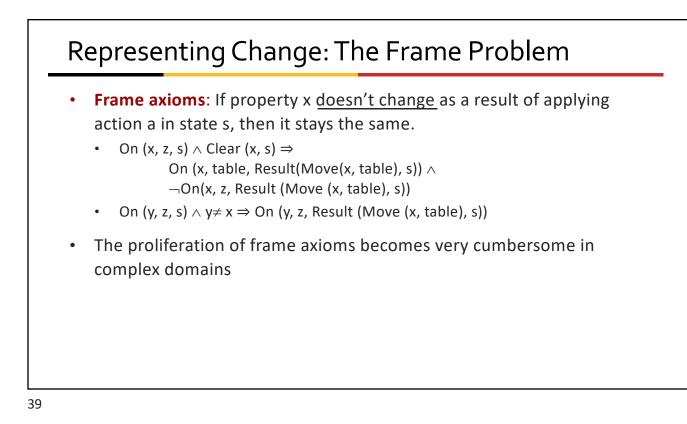
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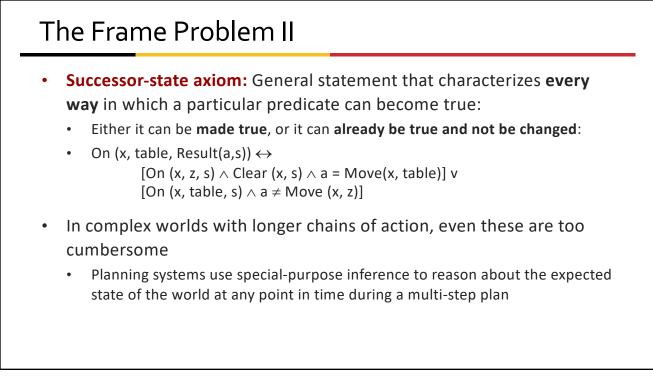
Frames: A Data Structure

- A frame divides knowledge into substructures by representing "stereotypical situations."
- Situations can be visual scenes, structures of physical objects, ...
- Useful for representing commonsense knowledge.

Slots	Fillers
publisher	Thomson
title	Expert Systems
author	Giarratano
edition	Third
year	1998
pages	600

Slot	Fillers	
name	computer	
specialization_of	a_kind_of machine	
types	(desktop, laptop,mainframe,super) if-added: Procedure ADD_COMPUTER	
speed	default: faster if-needed: Procedure FIND_SPEED	
location	(home,office,mobile)	
under_warranty	(yes, no)	

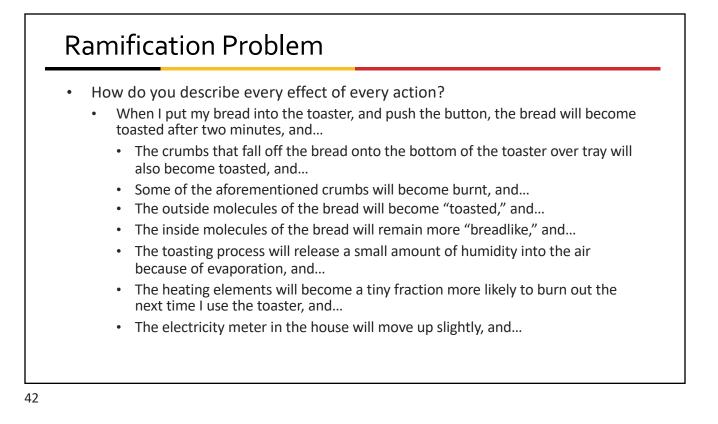




Qualification Problem

- Qualification problem: How can you possibly characterize every single effect of an action, or every single exception that might occur?
- When I put my bread into the toaster, and push the button, it will become toasted after two minutes, unless...
 - The toaster is broken, or...
 - The power is out, or...
 - I blow a fuse, or...
 - A neutron bomb explodes nearby and fries all electrical components, or...
 - A meteor strikes the earth, and the world we know it ceases to exist, or...





Knowledge Engineering!

- Modeling the "right" conditions and the "right" effects at the "right" level of abstraction is very difficult
- Knowledge engineering (creating and maintaining knowledge bases for intelligent reasoning) is a field
- Many researchers hope that automated knowledge acquisition and machine learning tools can fill the gap:
 - Our intelligent systems should be able to learn about the conditions and effects, just like we do.
 - Our intelligent systems should be able to learn when to pay attention to, or reason about, certain aspects of processes, depending on the context.

Preferences Among Actions

- A problem with the Wumpus world knowledge base: It's hard to decide which action is best!
 - Ex: to decide between a forward and a grab, axioms describing when it is okay to move would have to mention glitter.
- This is not modular!
- We can solve this problem by **separating facts about actions from facts about goals.**
- This way our agent can be reprogrammed just by asking it to achieve different goals.

Preferences Among Actions

- The first step is to describe the desirability of actions independent of each other.
- In doing this we will use a simple scale: actions can be Great, Good, Medium, Risky, or Deadly.
- Obviously, the agent should always do the best action it can find:
 - $(\forall a,s) \text{ Great}(a,s) \Rightarrow \text{Action}(a,s)$
 - $(\forall a,s) \text{ Good}(a,s) \land \neg(\exists b) \text{ Great}(b,s) \Rightarrow \text{Action}(a,s)$
 - $(\forall a,s) \text{ Medium}(a,s) \land (\neg(\exists b) \text{ Great}(b,s) \lor \text{Good}(b,s)) \Rightarrow \text{Action}(a,s)$

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Preferences Among Actions

- We use this action quality scale in the following way.
- Until it finds the gold, the basic strategy for our agent is:
 - Great actions include picking up the gold when found and climbing out of the cave with the gold.
 - Good actions include moving to a square that's OK and hasn't been visited yet.
 - Medium actions include moving to a square that is OK and has already been visited.
 - Risky actions include moving to a square that is not known to be deadly or OK.
 - Deadly actions are moving into a square that is known to have a pit or a Wumpus.

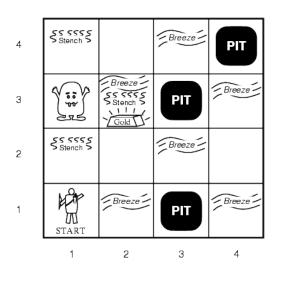
Goal-Based Agents

- Once the gold is found, it is necessary to change strategies. So now we need a new set of action values.
- We could encode this as a rule:
 - (∀s) Holding(Gold,s) ⇒ GoalLocation([1,1]),s)
- We must now decide how the agent will work out a sequence of actions to accomplish the goal.
- Three possible approaches are:
 - Inference: good versus wasteful solutions
 - Search: make a problem with operators and set of states
 - Planning: coming soon!

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An agent needs to make decisions!

- Where is there a pit?
- Where is there a wumpus?
- Should I fire my arrow?
- Where to explore next?
- Need to draw conclusions from knowledge in the knowledge base
- → Inference!





Chapter 9

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Review: English to FOL using quantifiers3. There is somebody who is loved by everyone. $\exists x \forall y (loves(y,x))$ - this person also loves themselves6. Frogs are green. $\forall x (frog(x) \Rightarrow green(x))$ - this is how we express rules10. A mechanic likes Bob. $\exists x (mech.(x) \land likes(x, Bob))$ - express existence of something• Usually:• Use AND with \exists (so we don't have a false antecedent)• Use IMPLIES with \forall (so we don't make too-strong claims)

Syntactic Ambiguity

- FOL provides many ways to represent the same thing.
- E.g., "Ball-5 is red."
 - HasColor(Ball-5, Red): Ball-5 and Red are objects related by HasColor.
 - **Red(Ball-5):** Red is a unary predicate applied to the Ball-5 object.
 - HasProperty(Ball-5, Color, Red): Ball-5, Color, and Red are objects related by HasProperty.
 - **ColorOf(Ball-5) = Red**: Ball-5 and Red are objects, and ColorOf() is a function.
 - HasColor(Ball-5(), Red()): Ball-5() and Red() are functions of zero arguments that both return an object, which objects are related by HasColor.
- This can GREATLY confuse a pattern-matching reasoner.



More choices to be made

- "For every food, there is a person who eats that food."
- [Use: Food(x), Person(y), Eats(y, x)]
 - $\forall x \exists y Food(x) \Rightarrow (Person(y) \land Eats(y, x))$
 - $\forall x \text{ Food}(x) \Rightarrow \exists y (Person(y) \land Eats(y, x))$
 - $\forall x \exists y \neg Food(x) \lor (Person(y) \land Eats(y, x))$
 - $\forall x \exists y (\neg Food(x) \lor Person(y)) \land (\neg Food(x) \lor Eats(y, x))$
 - $\forall x \exists y (Food(x) \Rightarrow Person(y)) \land (Food(x) \Rightarrow Eats(y, x))$
- Common Mistakes:
 - $\forall x \exists y (Food(x) \land Person(y)) \Rightarrow Eats(y, x)$
 - $\forall x \exists y Food(x) \land Person(y) \land Eats(y, x)$

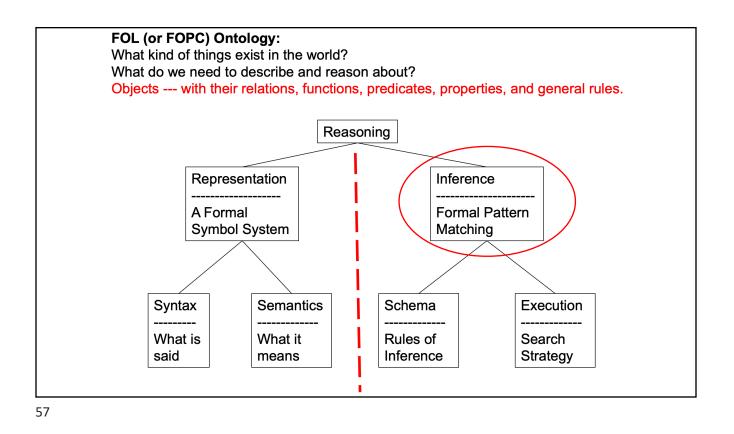
And yet more...

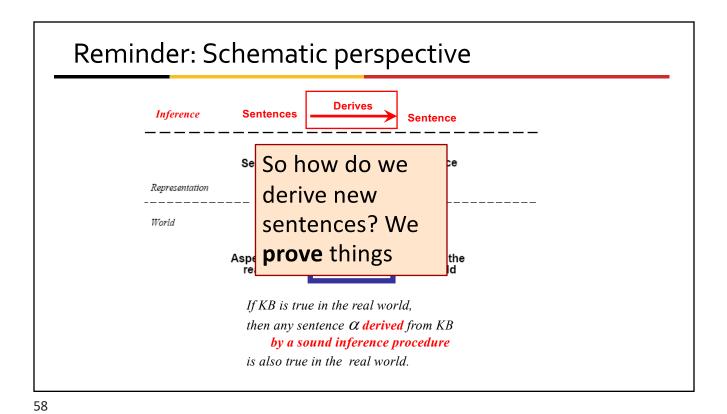
- "Every person eats some food."
- [Use: Person (x), Food (y), Eats(x, y)]
 - $\forall x \exists y \operatorname{Person}(x) \Rightarrow [\operatorname{Food}(y) \land \operatorname{Eats}(x, y)]$
 - $\forall x \operatorname{Person}(x) \Rightarrow \exists y [\operatorname{Food}(y) \land \operatorname{Eats}(x, y)]$
 - ∀x ∃y ¬Person(x) ∨ [Food(y) ∧ Eats(x, y)]
 - ∀x ∃y [¬Person(x) ∨ Food(y)] ∧ [¬Person(x) ∨ Eats(x, y)]
- Common Mistakes:
 - $\forall x \exists y [Person(x) \land Food(y)] \Rightarrow Eats(x, y)$
 - $\forall x \exists y Person(x) \land Food(y) \land Eats(x, y)$



Syntactic Ambiguity—Partial Solution

- FOL can be too expressive, can offer too many choices
- Likely confusion, especially for teams of Knowledge Engineers
- Different team members can make different representation choices
 - E.g., represent "Ball43 is Red." as:
 - a predicate (= verb)? E.g., "Red(Ball43)"?
 - an object (= noun)? E.g., "Red = Color(Ball43))" ?
 - a property (= adjective)? E.g., "HasProperty(Ball43, Red)"?
- Partial solution
 - An upon-agreed ontology that settles these questions
 - Ontology = what exists in the world & how it is represented
 - The Knowledge Engineering teams agrees upon an ontology BEFORE they begin encoding knowledge

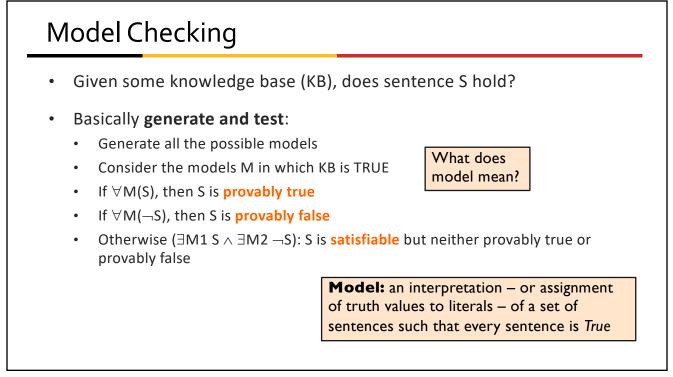


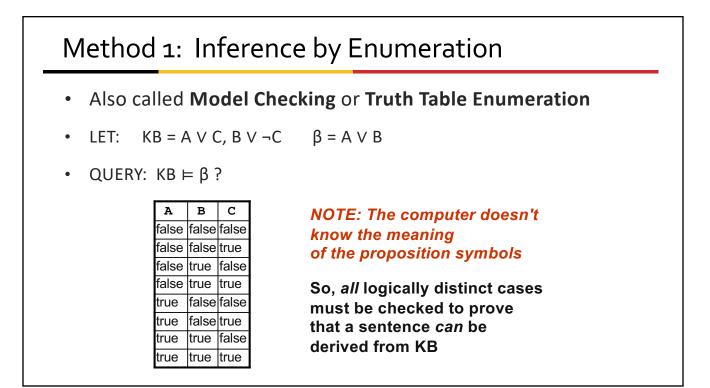


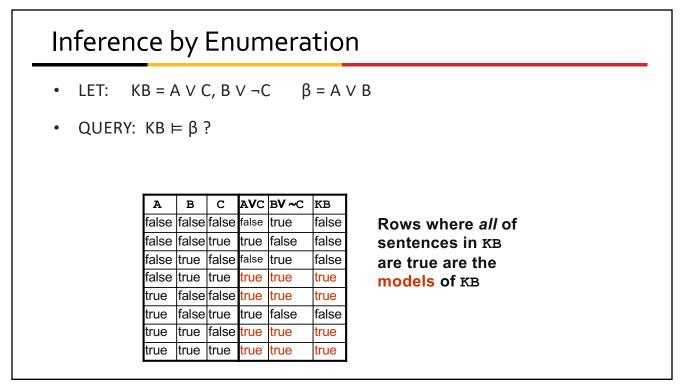
Proof methods

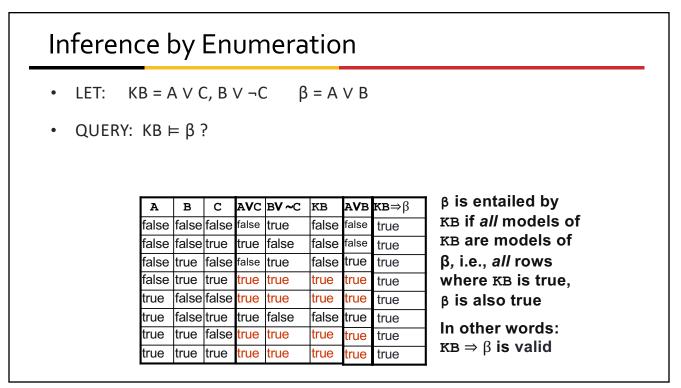
- Proof methods divide into (roughly) two kinds:
- Model checking:
 - Searching through truth assignments.
 - Improved backtracking: Davis-Putnam-Logemann-Loveland (DPLL)
 - Heuristic search in model space: Walksat
- Application of inference rules:
 - Legitimate (sound) generation of new sentences from old.
 - Forward & Backward chaining
 - Resolution KB is in Conjunctive Normal Form (CNF)





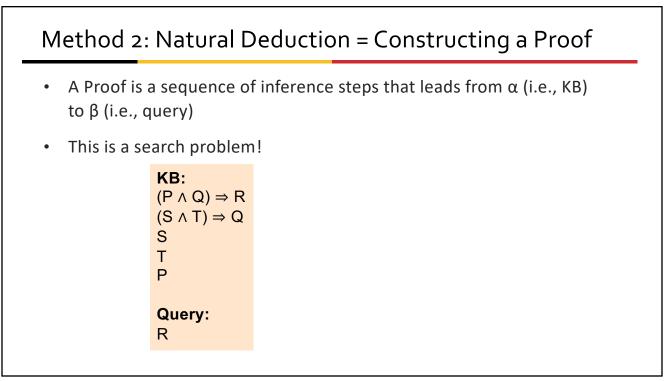






Inference by Enumeration

- Using inference by enumeration to build a complete truth table in order to tell if a sentence is entailed by KB is a complete inference algorithm for Propositional Logic
- But very slow: exponential time
- Imagine we had 5 literals... or 30, or hundreds, or millions



Proof by Natural Deduction					
KB:	1.	S	Premise (i.e., given sen	tence in KB)	
$(P \land Q) \Rightarrow R$ $(S \land T) \Rightarrow Q$	2.	Т	Premise		
(3 / 1) → Q S	3.	$S \wedge T$	Conjunction(1, 2) (And-Introduction)		
T P	4.	$(S\wedgeT)\RightarrowQ$	Premise		
	5.	Q	Modus Ponens(3, 4)	This is the	
Query: R	6.	Р	Premise	expected	
	7.	ΡΛQ	Conjunction(5, 6)	format of	
	8.	$(P\landQ)\RightarrowR$	Premise	proofs in the	
	9.	R	Modus Ponens(7, 8)	homework!	

Proof by Natural Deduction

- KB:
 - HaveLecture ⇔ (isTuesday ∨ isThursday)
 - ¬ HaveLecture
- Query:
 - ¬ isTuesday

Proof		
FIUUI	Contraposition:	
KB:	$(P\RightarrowQ)\equiv(\negQ\Rightarrow\negP)$	
1. HaveLecture ⇔ (isTuesday ∨ isThursday)		
2. ¬HaveLecture		
 (HaveLecture ⇒ (isTuesday ∨ isThursday)) ∧ ((isTuesday ∨ isThursday) ⇒ HaveLecture) 	iff elimination to 1	
4. (isTuesday ∨ isThursday) \Rightarrow HaveLecture	and-elimination to 3	
5. ¬HaveLecture \Rightarrow ¬(isTuesday ∨ isThursday)	contraposition to 4	
6. ¬(isTuesday ∨ isThursday)	Modus Ponens 2,5	
7. −isTuesday ∧ −isThursday	de Morgan to 6	
8. –isTuesday	and-elimination to 7	

Automating FOL Inference with Generalized Modus Ponens

Automated Inference for FOL

- Automated inference using FOL is harder than PL
 - Variables can take on an infinite number of possible values
 - From their domains, anyway
 - This is a reason to do careful KR!
 - So, potentially infinite ways to apply Universal Elimination
- Godel's Completeness Theorem says that FOL entailment is only semidecidable*
 - If a sentence is true given a set of axioms, can prove it
 - If the sentence is **false**, then there is no guarantee that a procedure will ever determine this
 - Inference may never halt

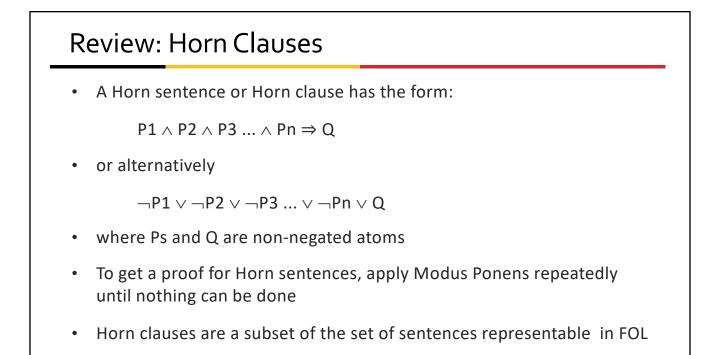
*The "halting problem"

Generalized Modus Ponens (GMP)

- Apply modus ponens reasoning to generalized rules
- Combines And-Introduction, Universal-Elimination, and Modus Ponens
 - From P(c) and Q(c) and $(\forall x)(P(x) \land Q(x)) \Rightarrow R(x)$ derive R(c)
- General case: Given
 - atomic sentences P₁, P₂, ..., P_N
 - implication sentence $(Q_1 \land Q_2 \land ... \land Q_N) \Rightarrow R$
 - Q₁, ..., Q_N and R are atomic sentences
 - **substitution** subst(θ, P_i) = subst(θ, Q_i) for i=1,...,N
 - Derive new sentence: subst(θ, R)



5 Derive new sentence: subst(θ, R) Substitutions subst(θ, α) denotes the result of applying a set of substitutions, defined by θ, to the sentence α A substitution list θ = {v₁/t₁, v₂/t₂, ..., v_n/t_n} means to replace all occurrences of variable symbol v_i by term t_i Substitutions are made in left-to-right order in the list subst({x/lceCream, y/Ziggy}, eats(y,x)) = eats(Ziggy, lceCream)



Horn Clauses II

- These are Horn clauses (special cases):
 - $P1 \land P2 \land ... Pn \Rightarrow Q$
 - $P1 \land P2 \land ... Pn \Rightarrow false$
 - true \Rightarrow Q
- These are not Horn clauses:
 - $p \lor q$ (all but one literal must be negated)
 - $(P \land Q) \Rightarrow (R \lor S)$ (non-literal after the implication)

Inference with Horn KBs

- If everything is Horn clauses, only 1 rule of inference needed
- Generalized Modus Ponens (GMP):

Given P, Q, and $(P \land Q) \Rightarrow R$, conclude R

• Written as:

$$\frac{P, Q, (P \land Q) \Rightarrow R}{R}$$

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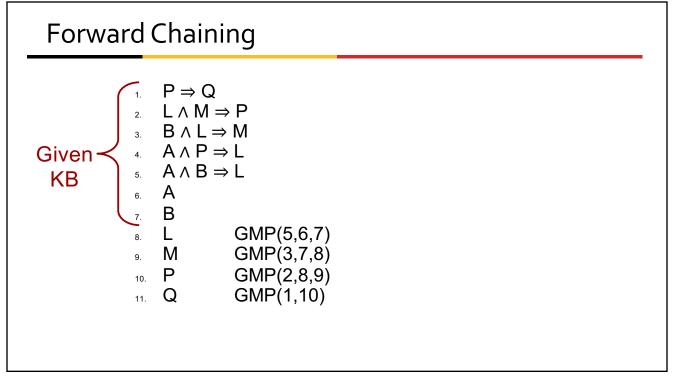
Method 3: Forward Chaining

- Proofs start with the given axioms/premises in KB, deriving new sentences using GMP until the goal/query sentence is derived
- This defines a **forward-chaining** inference procedure because it moves "forward" from the KB to the goal [eventually]
- Forward chaining with Horn clause KB is complete
 - A formal system is called complete with respect to a particular property if every formula having the property can be derived using that system, i.e. is one of its theorems;
 - Intuitively, a system is called complete if it can derive every formula that is true.

Forward Chaining

- "Apply" any rule whose premises are satisfied in the KB
- Add its conclusion to the KB until query is derived

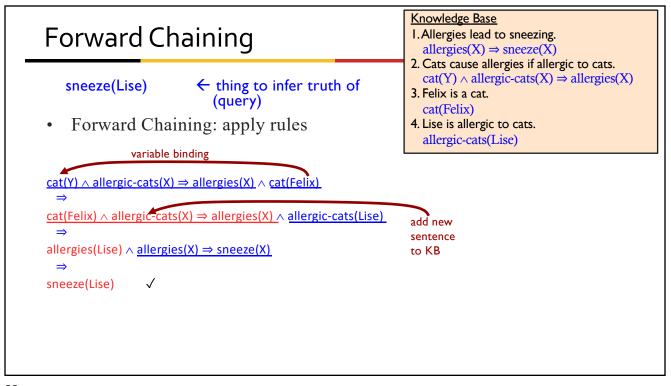
 $P \Rightarrow Q$ $L \land M \Rightarrow P$ $B \land L \Rightarrow M$ $A \land P \Rightarrow L$ $A \land B \Rightarrow L$ A Bquery: Q



Forward Chaining Example

• KB:

- allergies(X) \Rightarrow sneeze(X)
- $cat(Y) \land allergic-to-cats(X) \Rightarrow allergies(X)$
- cat(Felix)
- allergic-to-cats(Lise)
- Goal:
 - sneeze(Lise)



Forward Chaining Exercise

• Consider the following KB:

```
1. J \Rightarrow Q
```

2.	$A\landI\RightarrowJ$	8. E	(GMP 5,6,7)
3.	$E\landF\RightarrowI$	9. F	(GMP 4,7)
4.	$B \Rightarrow F$	10. I	(GMP 3,8,9)
5.	$A \land B \Rightarrow E$	11. J	(GMP 2,6,10)
6.	А		(GMP 1,11)
7.	В	12. Q	(UIVIF 1,11)

Prove Q. (Remember, you'll just use GMP over and over!)

```
• A, B, (A ∧ B) ⇒ C, ∴ C
```

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Method 4: Backward Chaining

- Forward chaining problem: can generate a lot of irrelevant conclusions
 - Search forward, start state = KB, goal test = state contains query
- Backward chaining
 - Work backwards from goal to premises
 - Find all implications of the form (...) \Rightarrow query
 - Prove all the premises of one of these implications
 - Avoid loops: check if new subgoal is already on the goal stack
 - Avoid repeated work: check if new subgoal
 - Has already been proved true, or
 - Has already failed

Backward Chaining

- Backward-chaining deduction using GMP
 - Complete for KBs containing only Horn clauses.
- Proofs:
 - Start with the goal query
 - Find rules with that conclusion
 - Prove each of the antecedents in the implication
- Keep going until you reach premises!

Avoid loops

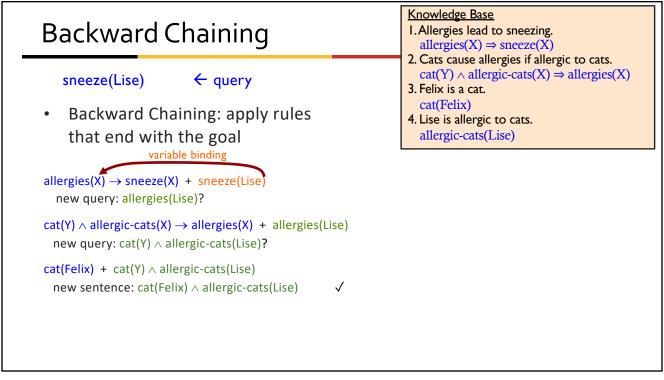
Is new subgoal already on goal stack?
Avoid repeated work: has subgoal
Already been proved true?

- Already failed?

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Backward Chaining Example

- KB:
 - allergies(X) \Rightarrow sneeze(X)
 - $cat(Y) \land allergic-to-cats(X) \Rightarrow allergies(X)$
 - cat(Felix)
 - allergic-to-cats(Lise)
- Goal:
 - sneeze(Lise)



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Forward vs. Backward Chaining

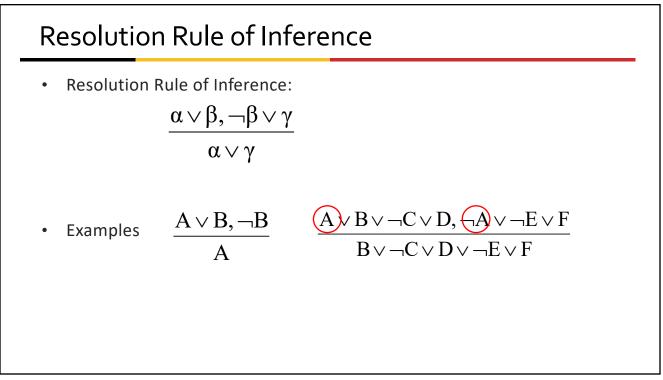
- FC is data-driven
 - Automatic, unconscious processing
 - E.g., object recognition, routine decisions
 - May do lots of work that is irrelevant to the goal
- BC is goal-driven, appropriate for problem-solving
 - Where are my keys? How do I get to my next class?
 - Complexity of BC can be much less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

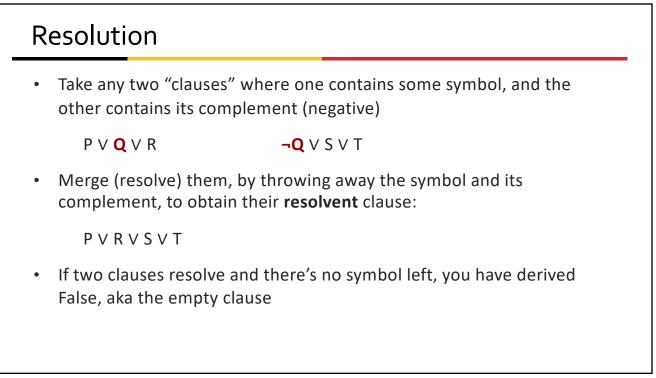
Completeness of GMP

- GMP (using forward or backward chaining) is complete for KBs that contain only Horn clauses
- It is *not* complete for simple KBs that contain non-Horn clauses
- The following KB entails that S(A) is true:
 - $(\forall x) P(x) \Rightarrow Q(x)$
 - $(\forall x) \neg P(x) \Rightarrow R(x)$
 - $(\forall x) Q(x) \Rightarrow S(x)$
 - $(\forall x) R(x) \Rightarrow S(x)$
- If we want to conclude S(A), with GMP we cannot, since the second one is not a Horn clause
- It is equivalent to $P(x) \vee R(x)$

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Automating FOL Inference with Resolution





Method 5: Resolution Refutation

- Show KB ⊨ α by proving that KB ∧ ¬α is unsatisifiable, i.e., deducing False from KB ∧ ¬α
- Your algorithm can use all the logical equivalences to derive new sentences, plus:
- **Resolution rule:** a *single* inference rule
 - Sound: only derives entailed sentences
 - Complete: can derive any entailed sentence
 - Resolution is refutation complete: if KB $\models \beta$, then KB $\land \neg \beta \vdash$ False
 - But the sentences need to be preprocessed into CNF
 - But all sentences can be converted into this form



Resolution Refutation Algorithm

- 1. Add negation of query to KB
- 2. Pick 2 sentences that haven't been used before and can be used with the Resolution Rule of inference
- 3. If none, halt and answer that the query is NOT entailed by KB
- 4. Compute resolvent and add it to KB
- 5. If False in KB
 - Then halt and answer that the query IS entailed by KB
 - Else Goto 2

Review: Converting to CNF

• Replace all ⇔ using biconditional elimination

• $\alpha \Leftrightarrow \beta \equiv (\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$

- Replace all \Rightarrow using implication elimination
 - $\alpha \Rightarrow \beta \equiv \neg \alpha \lor \beta$
- Move all negations inward using
 - double-negation elimination
 - $\neg(\neg \alpha) \equiv \alpha$
 - de Morgan's rule
 - $\neg(\alpha \lor \beta) \equiv \neg \alpha \land \neg \beta$
 - $\neg(\alpha \land \beta) \equiv \neg \alpha \lor \neg \beta$
- Apply distributivity of V over Λ
 - $\alpha \wedge (\beta \vee \gamma) \equiv (\alpha \wedge \beta) \vee (\alpha \wedge \gamma)$

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Resolution Refutation Steps

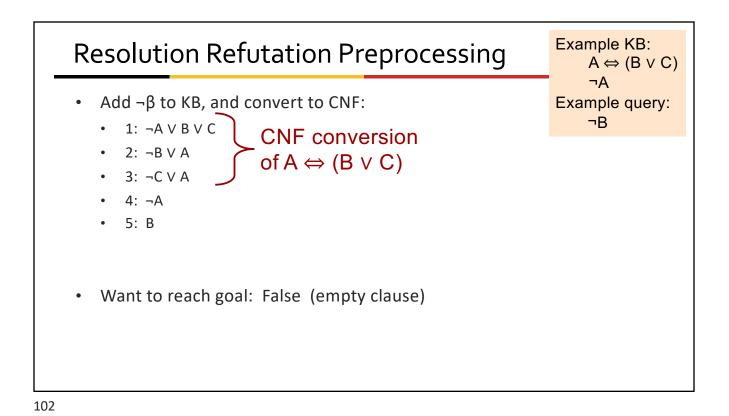
- Given KB and β (query)
- Add $\neg\beta$ to KB, and convert all sentences to CNF
- Show this leads to False (aka "empty clause"). Proof by contradiction
- Example KB:
 - $A \Leftrightarrow (B \lor C)$
 - ¬A
- Example query: ¬B

Result: something with clauses made up of ORs, separated by ANDs:

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

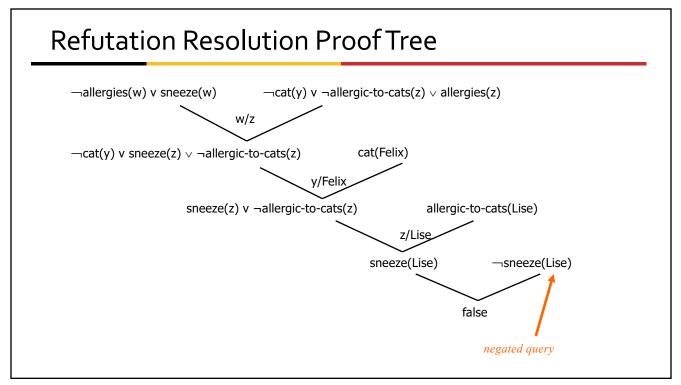
Review: Example Conversion to CNF

- Example: A \Leftrightarrow (B \vee C)
- Eliminate \Leftrightarrow by replacing $\alpha \Leftrightarrow \beta$ with $(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$.
 - $\bullet \quad = (\mathsf{A} \Rightarrow (\mathsf{B} \lor \mathsf{C})) \land ((\mathsf{B} \lor \mathsf{C}) \Rightarrow \mathsf{A})$
- 2. Eliminate \Rightarrow by replacing $\alpha \Rightarrow \beta$ with $\neg \alpha \lor \beta$ and simplify.
 - = $(\neg A \lor B \lor C) \land (\neg (B \lor C) \lor A)$
- 3. Move inwards using de Morgan's rules and simplify.
 - = $(\neg A \lor B \lor C) \land ((\neg B \land \neg C) \lor A)$
- 4. Apply distributive law (\land over \lor) and simplify.
 - = $(\neg A \lor B \lor C) \land (\neg B \lor A) \land (\neg C \lor A)$



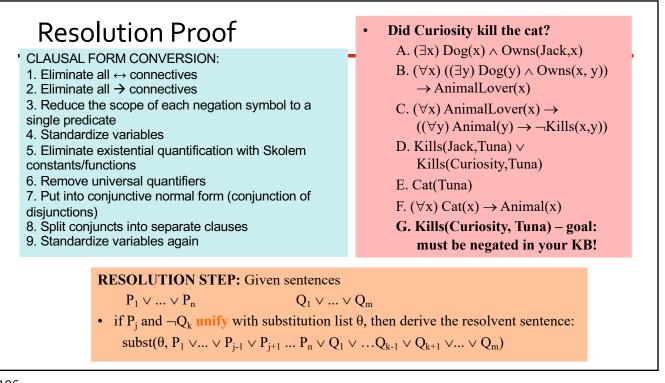
Resolution Refutation Example

- 1: ¬A V B V C
- 2: ¬B V A
- 3: ¬C V A
- 4: ¬A
- 5: B
- 6: A Resolve 2, 5
- 7: false/empty clause Resolve 6, 4



Exercise: Did Curiosity Kill the Cat?

- Jack owns a dog. Every dog owner is an animal lover. No animal lover kills an animal. Either Jack or Curiosity killed the cat, who is named Tuna. Did Curiosity kill the cat?
- These can be represented as follows:
 - A. $(\exists x) Dog(x) \land Owns(Jack,x)$
 - B. $(\forall x) ((\exists y) \text{Dog}(y) \land \text{Owns}(x, y)) \rightarrow \text{AnimalLover}(x)$
 - C. $(\forall x)$ AnimalLover $(x) \rightarrow ((\forall y) \text{ Animal}(y) \rightarrow \neg \text{Kills}(x,y))$
 - D. Kills(Jack,Tuna) V Kills(Curiosity,Tuna)
 - E. Cat(Tuna)
 - F. $(\forall x)$ Cat $(x) \rightarrow$ Animal(x)
 - G. Kills(Curiosity, Tuna)
 Query



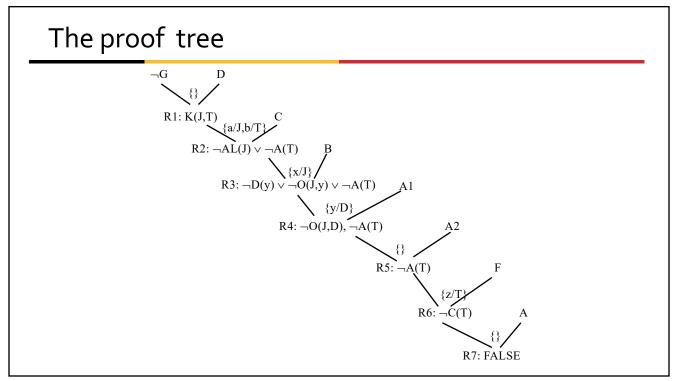
Steps

- Convert to clause form

 - A2. (Owns(Jack,D))
 - B. (¬Dog(y), ¬Owns(x, y), AnimalLover(x))
 - C. (¬AnimalLover(a), ¬Animal(b), ¬Kills(a,b))
 - D. (Kills(Jack,Tuna), Kills(Curiosity,Tuna))
 - E. Cat(Tuna)
 - F. (¬Cat(z), Animal(z))
- Add the negation of query:
 - ¬G: (¬Kills(Curiosity, Tuna))



The Resolution Refutation Proof			
• R1: ¬G, D, {}	(Kills(Jack, Tuna))		
 R2: R1, C, {a/Jack, b/Tuna} 	(~AnimalLover(Jack), ~Animal(Tuna))		
• R3: R2, B, {x/Jack}	(~Dog(y), ~Owns(Jack, y), ~Animal(Tuna))		
• R4: R3, A1, {y/D}	(~Owns(Jack, D), ~Animal(Tuna))		
• R5: R4, A2, {}	(~Animal(Tuna))		
• R6: R5, F, {z/Tuna}	(~Cat(Tuna))		
• R7: R6, E, {}	FALSE		



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Resolution Refutation

- Given a consistent set of axioms KB and goal sentence Q, show that KB = Q
- **Proof by contradiction:** Add \neg Q to KB and try to prove false.
 - i.e., (KB \vdash Q) \leftrightarrow (KB $\land \neg$ Q \vdash False)
- Resolution is refutation complete: it can establish that a given sentence Q is entailed by KB, but can't (in general) be used to generate all logical consequences of a set of sentences
 - Also, it cannot be used to prove that Q is not entailed by KB.

Efficiency of Resolution Refutation

- Run time can be exponential in the worst case
 - Often much faster
- Factoring: if a new clause contains duplicates of the same symbol, delete the duplicates
 - $P \vee R \vee P \vee T \equiv P \vee R \vee T$
- If a clause contains a symbol and its complement, the clause is a tautology and is useless; it can be thrown away
 - a1: (¬A V B V C)
 - a2: (¬B V A)
 - Resolvent of a1 and a2 is: $B \lor C \lor \neg B$
 - Which is valid, so throw it away



Resolution Theorem Proving as Search

- Resolution can be thought of as the bottom-up construction of a search tree, where the leaves are the clauses produced by KB and the negation of the goal
- When a pair of clauses generates a new resolvent clause, add a new node to the tree with arcs directed from the resolvent to the two parent clauses
- Resolution succeeds when a node containing the False clause is produced, becoming the root node of the tree
- A strategy is **complete** if its use guarantees that the empty clause (i.e., false) can be derived whenever it is entailed

Summary

- Logical agents apply inference to a knowledge base to derive new information and make decisions
- Basic concepts of logic:
 - syntax: formal structure of sentences
 - semantics: truth of sentences wrt models
 - entailment: necessary truth of one sentence given another
 - inference: deriving sentences from other sentences
 - soundness: derivations produce only entailed sentences
 - completeness: derivations can produce all entailed sentences
- Resolution is complete for propositional logic.
 Forward and backward chaining are linear-time, complete for Horn clauses
- Propositional logic lacks expressive power

Next Time

- Final bits of reasoning via inference
- Knowledge Representation
- Project work bring computers