

Lecture 6a: Rule Based System

CSCI 360

Introduction to Artificial Intelligence

USC

Here is where we are...

Week	30000D	30282R	Topics	Chapters
1	1/7 1/9	1/8 1/10	Intelligent Agents Problem Solving and Search	[Ch 1.1-1.4 and 2.1-2.4] [Ch 3.1-3.3]
2	1/14 1/16	1/15 1/17	Uninformed Search Heuristic Search (A*)	[Ch 3.3-3.4] [Ch 3.5]
3	1/21 1/23	1/22 1/24	Heuristic Functions Local Search	[Ch 3.6] [Ch 4.1-4.2]
	1/25		Project 1 Out	
4	1/28 1/30	1/29 1/31	Adversarial Search Knowledge Based Agents	[Ch 5.1-5.3] [Ch 7.1-7.3]
5	2/4 2/6	2/5 2/7	Propositional Logic Inference First-Order Logic	[Ch 7.4-7.5] [Ch 8.1-8.4]
	2/8 2/8		Project 1 Due Homework 1 Out	
6	2/11 2/13	2/12 2/14	Rule-Based Systems Search-Based Planning	[Ch 9.3-9.4] [Ch 10.1-10.3]
	2/15		Homework 1 Due	
7	2/18 2/20	2/19 2/21	SAT-Based Planning Knowledge Representation	[Ch 10.4] [Ch 12.1-12.5]
8	2/25 2/27	2/26 2/28	Midterm Review Midterm Exam	

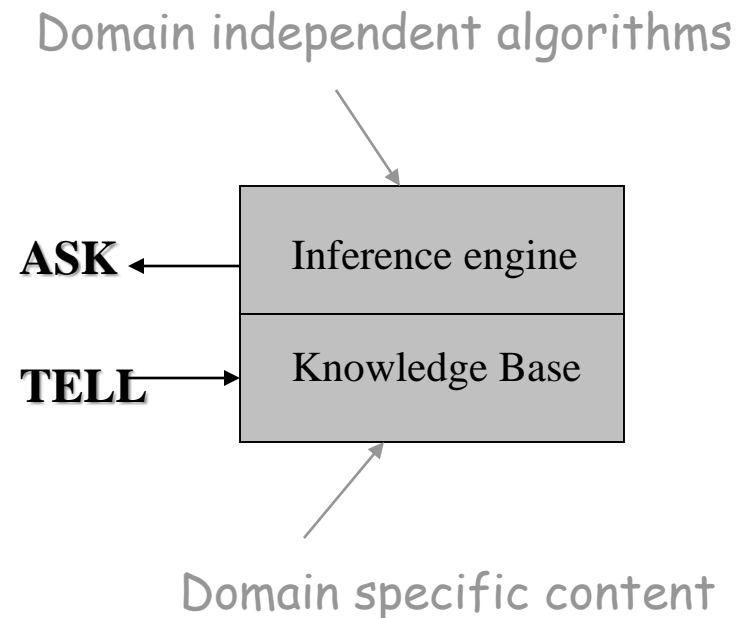


Outline

- What is AI?
- Problem-solving agent
 - Uninformed (DFS), informed (A^*), and local search
 - Adversarial search (minimax, alpha-beta pruning)
- **Knowledge-based agent**
 - Propositional Logic
 - First Order Logic (FOL)
 - **Automated Reasoning in FOL**

Recap: *Logic for knowledge representation*

- **Logic** as a language for knowledge representation
 - Propositional logic (Boolean)
 - First-order logic (FOL)



- Advantage
 - Can combine and recombine information to suit many purposes

Recap: *Propositional Logic: checking entailment*

- Two methods
 - Method#1: Based on enumeration (model checking)
 - Method#2: Based on inference rules (theorem proving)
- Enumerate all models and check if “ α is true in all models in which KB is true”

$$M(KB) \subseteq M(\alpha).$$

Recap: *PL*: Checking entailment (example)

$KB \models \alpha_1$

$KB \not\models \alpha_2$

α_1 = “There is no pit in [1,2].”

α_2 = “There is no pit in [2,2].”

$R_1 : \neg P_{1,1} .$

$R_2 : B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1}) .$

$R_3 : B_{2,1} \Leftrightarrow (P_{1,1} \vee P_{2,2} \vee P_{3,1}) .$

$R_4 : \neg B_{1,1} .$

$R_5 : B_{2,1} .$

KB

$B_{1,1}$	$B_{2,1}$	$P_{1,1}$	$P_{1,2}$	$P_{2,1}$	$P_{2,2}$	$P_{3,1}$	R_1	R_2	R_3	R_4	R_5	KB
false	false	false	false	false	false	false	true	true	true	true	false	false
false	false	false	false	false	false	true	true	true	false	true	false	false
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
false	true	false	false	false	false	false	true	true	false	true	true	false
false	true	false	false	false	false	true	true	true	true	true	true	true
false	true	false	false	false	true	false	true	true	true	true	true	true
false	true	false	false	true	false	false	true	false	false	true	true	false
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
true	true	true	true	true	true	true	false	true	true	false	true	false

Recap: *PL: Checking entailment (inference)*

- Two methods

- Method#1: Based on enumeration (model checking)
- Method#2: Based on inference rules (theorem proving)

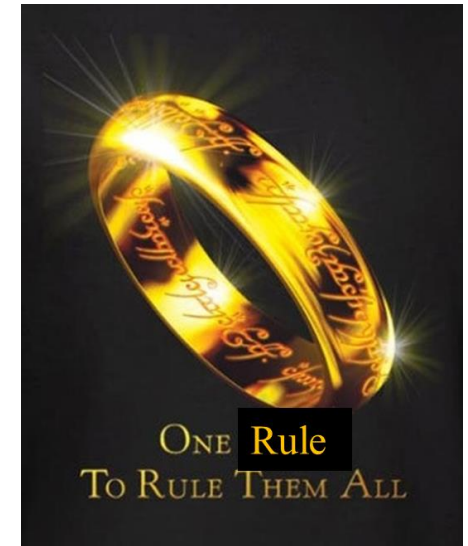
- Logical equivalences

$$\begin{aligned}(\alpha \wedge \beta) &\equiv (\beta \wedge \alpha) && \text{commutativity of } \wedge \\(\alpha \vee \beta) &\equiv (\beta \vee \alpha) && \text{commutativity of } \vee \\((\alpha \wedge \beta) \wedge \gamma) &\equiv (\alpha \wedge (\beta \wedge \gamma)) && \text{associativity of } \wedge \\((\alpha \vee \beta) \vee \gamma) &\equiv (\alpha \vee (\beta \vee \gamma)) && \text{associativity of } \vee \\\neg(\neg\alpha) &\equiv \alpha && \text{double-negation elimination} \\(\alpha \Rightarrow \beta) &\equiv (\neg\beta \Rightarrow \neg\alpha) && \text{contraposition} \\(\alpha \Rightarrow \beta) &\equiv (\neg\alpha \vee \beta) && \text{implication elimination} \\(\alpha \Leftrightarrow \beta) &\equiv ((\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)) && \text{biconditional elimination} \\\neg(\alpha \wedge \beta) &\equiv (\neg\alpha \vee \neg\beta) && \text{De Morgan} \\\neg(\alpha \vee \beta) &\equiv (\neg\alpha \wedge \neg\beta) && \text{De Morgan} \\(\alpha \wedge (\beta \vee \gamma)) &\equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) && \text{distributivity of } \wedge \text{ over } \vee \\(\alpha \vee (\beta \wedge \gamma)) &\equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) && \text{distributivity of } \vee \text{ over } \wedge\end{aligned}$$

Recap: *PL: Resolution rule*

- Resolvent
- Pivot variable (x)

$$(C \vee x) \wedge (D \vee \neg x) \rightarrow (C \vee D)$$



- **Example:** general case (= *transitivity of implication*)

$$\frac{P_{1,1} \vee P_{3,1}, \quad \neg P_{1,1} \vee \neg P_{2,2}}{P_{3,1} \vee \neg P_{2,2}}$$

$$\begin{aligned} & (\neg C \rightarrow x) \wedge (x \rightarrow D) \\ &= (\neg C \rightarrow D) \\ &= (C \vee D) \end{aligned}$$

Recap: *Why first-order logic (FOL)?*

- Propositional logic (PL) is limited because it only makes the **ontological commitment** that a world consists of **facts**
 - Facts: propositions that are either *true* or *false*
 - “*Don’t go forward if the Wumpus is in front of you*” takes 64 rules

Language	Ontological Commitment (What exists in the world)
Propositional logic	facts
First-order logic	facts, objects, relations
Temporal logic	facts, objects, relations, times
Probability theory	facts
Fuzzy logic	facts with degree of truth $\in [0, 1]$

Recap: *First-order logic (FOL)*

- Ontological commitments:
 - **Objects:** Wheel, door, body, engine, seat, car, passenger, driver
 - **Functions:** ColorOf(car)
 - **Relations:** Inside(car, passenger), Beside(driver, passenger)
 - **Properties:** IsOpen(door), IsOn(engine)



CarA



CarB

Function: ColorOf(**CarA**) = BLACK

ColorOf(**CarB**) = BLUE

Relation: ColorOfCar(**CarA**,BLACK) = True
 ColorOfCar(**CarA**,BLUE) = False

ColorOfCar(**CarB**,BLACK) = False
ColorOfCar(**CarB**,BLUE) = True

Property: IsBlackCar(**CarA**) = True
 IsBlueCar(**CarA**) = False

IsBlackCar(**CarB**) = False
IsBlueCar(**CarB**) = True

Recap: *Universal quantification (for all): \forall*

\forall <variables> <**sentence**>

Example

- “Everyone in the cs360 class is smart”:

$$\forall x \quad \text{In (cs360, } x) \Rightarrow \text{Smart (} x)$$

- $\forall x$ means the conjunction of all instantiations of x

$$(\text{In (cs360, Markus)} \Rightarrow \text{Smart (Markus)}) \wedge$$

$$(\text{In (cs360, Dora)} \Rightarrow \text{Smart (Dora)}) \wedge$$

...

$$(\text{In (cs360, Hao)} \Rightarrow \text{Smart (Hao)})$$

Recap: *Existential quantification (there exists): \exists*

\exists <variables> <sentence>

Example

- “Someone in the cs360 class is smart”:

$\exists x \text{ In (cs360, } x) \wedge \text{Smart (} x)$

- $\exists x$ represents the disjunction of all instantiations of x

$\text{In (cs360, Markus) } \wedge \text{Smart (Markus) } \vee$

$\text{In (cs360, Dora) } \wedge \text{Smart (Dora) } \vee$

...

$\text{In (cs360, Hao) } \wedge \text{Smart (Hao)}$

Using FOL Knowledge Base (*Wumpus*)

$$\forall x, y, a, b \text{ } Adjacent([x, y], [a, b]) \Leftrightarrow$$

Using FOL Knowledge Base (*Wumpus*)

$$\forall x, y, a, b \text{ } Adjacent([x, y], [a, b]) \Leftrightarrow \\ (x = a \wedge (y = b - 1 \vee y = b + 1)) \vee (y = b \wedge (x = a - 1 \vee x = a + 1))$$

Squares are breezy near a pit:

Diagnostic rule—infer cause from effect

$$\forall y \text{ } Breezy(y) \Rightarrow \exists x \text{ } Pit(x) \wedge Adjacent(x, y)$$

Using FOL Knowledge Base (*Wumpus*)

$$\forall x, y, a, b \text{ } Adjacent([x, y], [a, b]) \Leftrightarrow \\ (x = a \wedge (y = b - 1 \vee y = b + 1)) \vee (y = b \wedge (x = a - 1 \vee x = a + 1))$$

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$$\forall y \text{ } Breezy(y) \Rightarrow \exists x \text{ } Pit(x) \wedge Adjacent(x, y)$$

Causal rule—infer effect from cause

$$\forall x, y \text{ } Pit(x) \wedge Adjacent(x, y) \Rightarrow Breezy(y)$$

Using FOL Knowledge Base (*Wumpus*)

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Neither of these is complete—e.g., the causal rule doesn't say whether squares far away from pits can be breezy

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Definition for the *Breezy* predicate:

$$\forall y \text{ } Breezy(y) \Leftrightarrow [\exists x \text{ } Pit(x) \wedge Adjacent(x, y)]$$

Equality

$term_1 = term_2$ is true under a given interpretation
if and only if $term_1$ and $term_2$ refer to the same object

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E.g., definition of (full) *Sibling* in terms of *Parent*:

$$\forall x, y \text{ Sibling}(x, y) \Leftrightarrow [\exists m, f \\ \text{Parent}(m, x) \wedge \text{Parent}(f, x) \wedge \text{Parent}(m, y) \wedge \text{Parent}(f, y)]$$

Equality

$term_1 = term_2$ is true under a given interpretation
if and only if $term_1$ and $term_2$ refer to the same object

E.g., definition of (full) *Sibling* in terms of *Parent*:

$$\forall x, y \text{ Sibling}(x, y) \Leftrightarrow [\neg(x = y) \wedge \exists m, f \neg(m = f) \wedge \\ \text{Parent}(m, x) \wedge \text{Parent}(f, x) \wedge \text{Parent}(m, y) \wedge \text{Parent}(f, y)]$$

Higher-order logic?

- First-order logic allows us to quantify over **objects**.
- Higher-order logic allows us to quantify over **objects**, **relations**, and **functions**.

e.g., “two objects are equal if and only if all **properties** applied to them are equivalent”:

$$\forall x,y \quad (x=y) \Leftrightarrow (\forall p \quad p(x) \Leftrightarrow p(y))$$

- Higher-order logics are **more expressive**; however, it is not clear (yet) *how to effectively reason about* sentences in higher-order logic.

Outline for Today

- What is AI?
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 - Adversarial search (minimax, alpha-beta pruning)
- **Knowledge-based agent**
 - Propositional Logic
 - First Order Logic (FOL)
 - **Reasoning in FOL**
 - Substitution
 - Unification
 - Chaining (forward and backward)
 - Resolution

Substitution

- Assume that (KB) has the following axiom:

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

- How to infer any of the following sentences?

$$\text{King}(\text{John}) \wedge \text{Greedy}(\text{John}) \Rightarrow \text{Evil}(\text{John})$$

$$\text{King}(\text{Richard}) \wedge \text{Greedy}(\text{Richard}) \Rightarrow \text{Evil}(\text{Richard})$$

$$\text{King}(\text{Father}(\text{John})) \wedge \text{Greedy}(\text{Father}(\text{John})) \Rightarrow \text{Evil}(\text{Father}(\text{John})) .$$

\vdots

Substitution

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$$\text{King}(\text{Father}(\text{John})) \wedge \text{Greedy}(\text{Father}(\text{John})) \Rightarrow \text{Evil}(\text{Father}(\text{John})) .$$

⋮

- Replacing a variable by a ground term
 - i.e., a term without variables (constant, or a function applied to a constant, or a function applied to another ground term)

$$\{x/\text{John}\}, \{x/\text{Richard}\}, \text{ and } \{x/\text{Father}(\text{John})\}$$

Universal Elimination

- Substitute a **variable** with a **ground term** (one without variables)

$$\frac{\forall v \ \alpha}{\text{SUBST}(\{v/g\}, \alpha)}$$

- Example:**

$$\forall x \ King(x) \wedge Greedy(x) \Rightarrow Evil(x) .$$

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

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

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

\vdots

Existential Elimination

- Substitute a **variable** with a **single, new, constant symbol**

$$\frac{\exists v \ \alpha}{\text{SUBST}(\{v/k\}, \alpha)}$$

- Example:**

$$\exists x \ \text{Crown}(x) \wedge \text{OnHead}(x, \text{John})$$

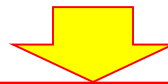
$$\text{Crown}(C_1) \wedge \text{OnHead}(C_1, \text{John})$$

as long as C_1 does not appear elsewhere in the knowledge base.

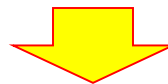
Reducing FOL to PL for inference

- Apply both “existential” and “universal” eliminations and then discard the quantified sentences
 - Universal elimination using all **ground-term substitutions** from the vocabulary of the knowledge base

$\forall x \text{ King}(x) \wedge \text{Greedy}(x) \Rightarrow \text{Evil}(x)$
 $\text{King}(\text{John})$
 $\text{Greedy}(\text{John})$
 $\text{Brother}(\text{Richard}, \text{John}) .$



$\{x/\text{John}\}$ and $\{x/\text{Richard}\}$

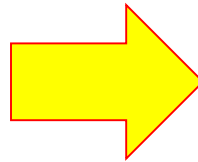


$\text{King}(\text{John}) \wedge \text{Greedy}(\text{John}) \Rightarrow \text{Evil}(\text{John})$
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Reducing FOL to PL for inference

- Apply both “existential” and “universal” eliminations and then discard the quantified sentences
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KB (in FOL)



KB (in PL)

$\forall x \text{ King}(x) \wedge \text{Greedy}(x) \Rightarrow \text{Evil}(x)$
 $\text{King}(\text{John})$
 $\text{Greedy}(\text{John})$
 $\text{Brother}(\text{Richard}, \text{John}) .$

~~$\forall x \text{ King}(x) \wedge \text{Greedy}(x) \Rightarrow \text{Evil}(x)$~~
 $\text{King}(\text{John})$
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 $\text{Brother}(\text{Richard}, \text{John}) .$
 $\text{King}(\text{John}) \wedge \text{Greedy}(\text{John}) \Rightarrow \text{Evil}(\text{John})$
 $\text{King}(\text{Richard}) \wedge \text{Greedy}(\text{Richard}) \Rightarrow \text{Evil}(\text{Richard})$

Sound and Complete?

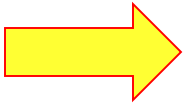
- Every FOL knowledge base (KB) and query (α) can be reduced to propositional (PL) logic to preserve entailment
 - Since PL inference is both “sound” and “complete”, it seems that we will have a similar procedure for FOL...
- **Not so fast!**
- When KB includes a function symbol, the set of possible ground-term substitution is **infinite!**
 - $Father(John)$, $Father(Father(John))$, $Father(Father(Father(John)))$,...

Herbrand's Theorem (1930)

- If a sentence is entailed by the KB in FOL, there is a proof involving just a **finite subset** of the propositionalized KB.
- **Corollary:** we can always find that proof, by iteratively deepening the function's nesting depth
 - Iteration 1: John, ...
 - Iteration 2: Father(John), ...
 - Iteration 3: Father(Father(John)), ...
 - Iteration 4: Father(Father(Father(John))), ...
- **Complete:** Any entailed sentence can be proved
- **Unsound:** Can't prove that a sentence is not entailed

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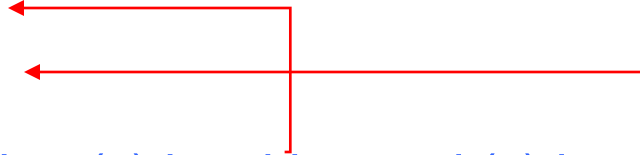
FOL inference rule

- Reducing FOL to PL is rather inefficient; sometimes, direct inference seems more intuitive

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

King(John)

Greedy(John)



- Step 1: Find some (x) such that (x) is a king and (x) is greedy

substitution $\theta = \{x / \text{John}\}$

- Step 2: Infer that this particular (x) is evil.

Evil(John)

FOL inference rule

- Reducing FOL to PL is rather inefficient; sometimes, direct inference seems more intuitive

$$\begin{array}{l} \forall x \text{ King}(x) \wedge \text{Greedy}(x) \Rightarrow \text{Evil}(x) \\ \text{King}(\text{John}) \\ \text{Greedy}(\text{John}) \end{array} \quad \forall y \text{ Greedy}(y)$$

- Step 1: Find some (x) such that (x) is a king and (x) is greedy

~~substitution $\theta = \{x/\text{John}\}$~~ substitution $\{x/\text{John}, y/\text{John}\}$

- Step 2: Infer that this particular (x) is evil.

$$\text{Evil}(\text{John})$$

Generalized Modus Ponens

- There is a **substitution** such that

$$\text{SUBST}(\theta, p_i') = \text{SUBST}(\theta, p_i), \text{ for all } i,$$

$$\frac{p_1', p_2', \dots, p_n', (p_1 \wedge p_2 \wedge \dots \wedge p_n \Rightarrow q)}{\text{SUBST}(\theta, q)}$$

- Example

$\text{King}(\text{John})$

$\forall y \text{ Greedy}(y)$

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

p_1' is $\text{King}(\text{John})$

p_2' is $\text{Greedy}(y)$

p_1 is $\text{King}(x)$

p_2 is $\text{Greedy}(x)$

Generalized Modus Ponens

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$\text{King}(\text{John})$

$\forall y \text{ Greedy}(y)$

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

p_1' is $\text{King}(\text{John})$

p_2' is $\text{Greedy}(y)$

θ is $\{x/\text{John}, y/\text{John}\}$

p_1 is $\text{King}(x)$

p_2 is $\text{Greedy}(x)$

q is $\text{Evil}(x)$

Generalized Modus Ponens

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$$\frac{p_1', \quad p_2', \quad \dots, \quad p_n', \quad (p_1 \wedge p_2 \wedge \dots \wedge p_n \Rightarrow q)}{\text{SUBST}(\theta, q)}$$

- Example

$\text{King}(\text{John})$

$\forall y \text{ Greedy}(y)$

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

p_1' is $\text{King}(\text{John})$

p_1 is $\text{King}(x)$

p_2' is $\text{Greedy}(y)$

p_2 is $\text{Greedy}(x)$

θ is $\{x/\text{John}, y/\text{John}\}$

q is $\text{Evil}(x)$

$\text{SUBST}(\theta, q)$ is $\text{Evil}(\text{John})$.

GMP example

Bob is a buffalo

Pat is a pig

Buffaloes outrun pigs

Bob outruns Pat

1. $Buffalo(Bob)$

2. $Pig(Pat)$

3. $\forall x, y \text{ } Buffalo(x) \wedge Pig(y) \Rightarrow Faster(x, y)$

GMP example

Bob is a buffalo	1. $Buffalo(Bob)$
Pat is a pig	2. $Pig(Pat)$
Buffaloes outrun pigs	3. $\forall x, y \text{ } Buffalo(x) \wedge Pig(y) \Rightarrow Faster(x, y)$
Bob outruns Pat	
Al 1 & 2	4. $Buffalo(Bob) \wedge Pig(Pat)$

GMP example

Bob is a buffalo	1. $Buffalo(Bob)$
Pat is a pig	2. $Pig(Pat)$
Buffaloes outrun pigs	3. $\forall x, y \text{ } Buffalo(x) \wedge Pig(y) \Rightarrow Faster(x, y)$
Bob outruns Pat	
AI 1 & 2	4. $Buffalo(Bob) \wedge Pig(Pat)$
UE 3, $\{x/Bob, y/Pat\}$	5. $Buffalo(Bob) \wedge Pig(Pat) \Rightarrow Faster(Bob, Pat)$

GMP example

Bob is a buffalo Pat is a pig Buffaloes outrun pigs Bob outruns Pat	1. $Buffalo(Bob)$ 2. $Pig(Pat)$ 3. $\forall x, y \text{ } Buffalo(x) \wedge Pig(y) \Rightarrow Faster(x, y)$
AI 1 & 2	4. $Buffalo(Bob) \wedge Pig(Pat)$
UE 3, $\{x/Bob, y/Pat\}$	5. $Buffalo(Bob) \wedge Pig(Pat) \Rightarrow Faster(Bob, Pat)$
MP 6 & 7	6. $Faster(Bob, Pat)$

Unification

- **Finding substitutions** that make different logical formulas look identical

$$\text{UNIFY}(p, q) = \theta \text{ where } \text{SUBST}(\theta, p) = \text{SUBST}(\theta, q)$$

- **Example**

$$\text{UNIFY}(\text{Knows}(\text{John}, x), \text{Knows}(\text{John}, \text{Jane})) = \{x/\text{Jane}\}$$

$$\text{UNIFY}(\text{Knows}(\text{John}, x), \text{Knows}(y, \text{Bill})) = \{x/\text{Bill}, y/\text{John}\}$$

$$\text{UNIFY}(\text{Knows}(\text{John}, x), \text{Knows}(y, \text{Mother}(y))) = \{y/\text{John}, x/\text{Mother}(\text{John})\}$$

$$\text{UNIFY}(\text{Knows}(\text{John}, x), \text{Knows}(x, \text{Elizabeth})) = \text{fail} .$$

Unification example

1 – unify(P(a,X), P(a,b))

2 – unify(P(a,X), P(Y,b))

3 – unify(P(a,X), P(Y,f(a)))

4 – unify(P(a,X), P(X,b))

VARIABLE term
 ↓ ↓

$\sigma = \{ X / b \}$

$\sigma = \{ Y/a, X/b \}$

$\sigma = \{ Y/a, X/f(a) \}$

$\sigma = \mathbf{failure}$

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 - **Chaining (forward and backward)**
 - Resolution

FOL definite clause

- **Universally quantified** clause in which exactly one literal is a positive literal

- **Example:**

$$\begin{aligned} & King(x) \wedge Greedy(x) \Rightarrow Evil(x) . \\ & King(John) . \\ & Greedy(y) . \end{aligned}$$

- We omit the “universal quantifier” in FOL definite clauses, but you should assume that, implicitly, all variables are universally quantified

FOL definite clause (*what about \exists ?*)

- Apply “**Existential Instantiation**” to make sure the KB has only universally quantified clauses

$$\exists x \text{ Owns}(\text{Nono}, x) \wedge \text{Missile}(x)$$



$$\text{Owns}(\text{Nono}, M_1)$$

$$\text{Missile}(M_1)$$

FOL forward chaining

- When a new fact (p) is added to the (KB)
 - For each rule such that (p) unifies with a premise
 - If the other premises are known, then
 - add the conclusion to the (KB) and continue chaining
- Forward chaining is “data-driven”
 - E.g., inferring facts from percepts

FOL forward chaining (example)

Add facts 1, 2, 3, 4, 5, 7 in turn.

Number in [] = unification literal; \checkmark indicates rule firing

FOL forward chaining (example)

Add facts 1, 2, 3, 4, 5, 7 in turn.

Number in [] = unification literal; \checkmark indicates rule firing

1. $Buffalo(x) \wedge Pig(y) \Rightarrow Faster(x, y)$

2. $Pig(y) \wedge Slug(z) \Rightarrow Faster(y, z)$

3. $Faster(x, y) \wedge Faster(y, z) \Rightarrow Faster(x, z)$

4. $Buffalo(Bob)$ [1a, \times]

5. $Pig(Pat)$ [1b, \checkmark] \rightarrow 6. $Faster(Bob, Pat)$ [3a, \times], [3b, \times]
[2a, \times]

7. $Slug(Steve)$ [2b, \checkmark]

\rightarrow 8. $Faster(Pat, Steve)$ [3a, \times], [3b, \checkmark]

\rightarrow 9. $Faster(Bob, Steve)$ [3a, \times], [3b, \times]

Fixed-point reached!

Properties of FOL forward chaining

- **Sound:** **Yes**, because every inference is an application of Generalized Modus Ponens
- **Complete:** **Yes**, it answers every query whose answer is entailed by the KB of FOL definite clauses

Question: *What if the query is not entailed by the KB?*

The algorithm may fail to terminate...

$$\begin{array}{l} \text{NatNum}(0) \\ \forall n \text{ NatNum}(n) \Rightarrow \text{NatNum}(S(n)) \end{array}$$

FOL backward chaining

- When a query (q) is asked
 - If a matching fact (q') is known in the (KB), return the unifier
 - For each rule whose consequence (q') matches (q)
 - Attempt to prove each of its premises, by backward chaining

FOL backward chaining (example)

1. $Pig(y) \wedge Slug(z) \Rightarrow Faster(y, z)$

2. $Slimy(z) \wedge Creeps(z) \Rightarrow Slug(z)$

3. $Pig(Pat)$

4. $Slimy(Steve)$

5. $Creeps(Steve)$

FOL backward chaining (example)

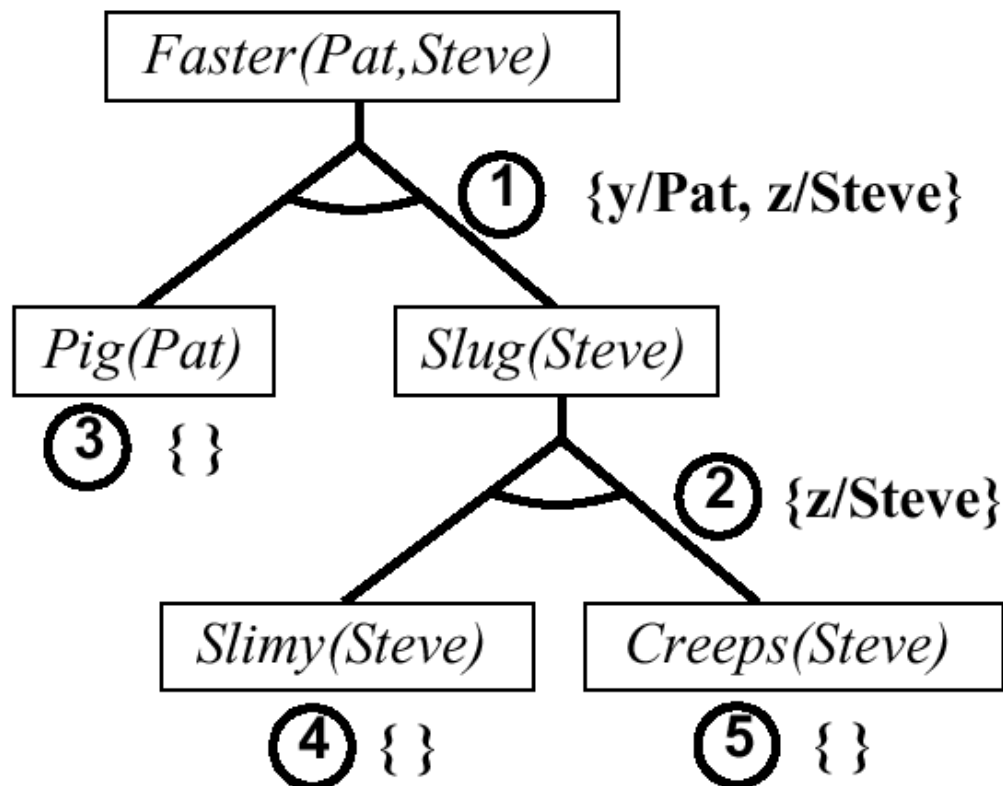
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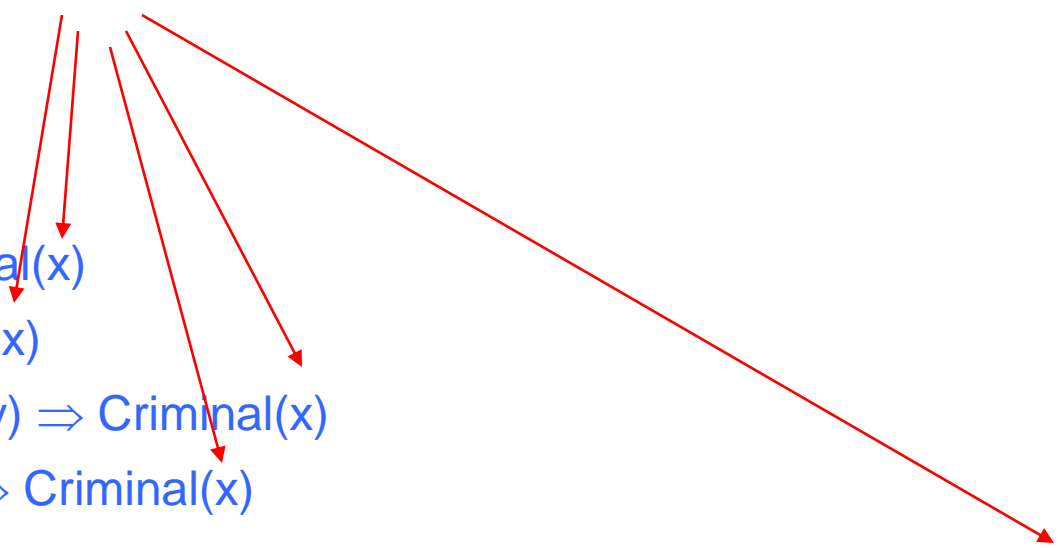


Backward Chaining

- Question: Has **Reality Man** done anything **criminal**?

- Criminal(Reality Man)

- Possible answers:

- Steal(x, y) \Rightarrow Criminal(x)
 - Kill(x, y) \Rightarrow Criminal(x)
 - Grow(x, y) \wedge Illegal(y) \Rightarrow Criminal(x)
 - HaveSillyName(x) \Rightarrow Criminal(x)
 - Programmer(x) \wedge Emulator(y) \wedge People(z) \wedge Provide(x,z,y) \Rightarrow Criminal(x)
- 

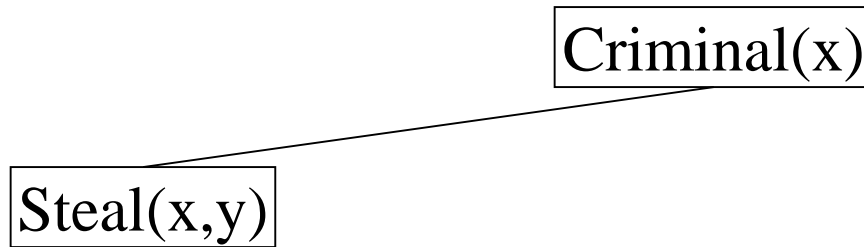
Backward Chaining

- Question: Has Reality Man done anything criminal?

Criminal(x)

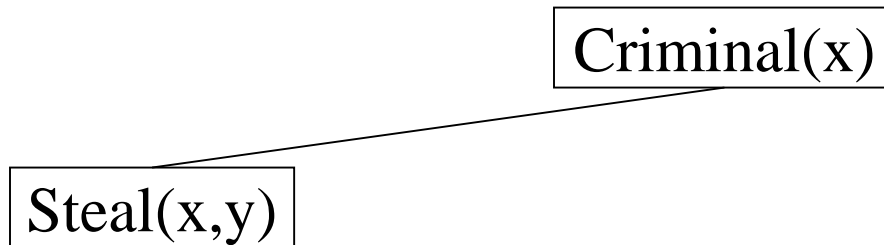
Backward Chaining

- Question: Has Reality Man done anything criminal?



Backward Chaining

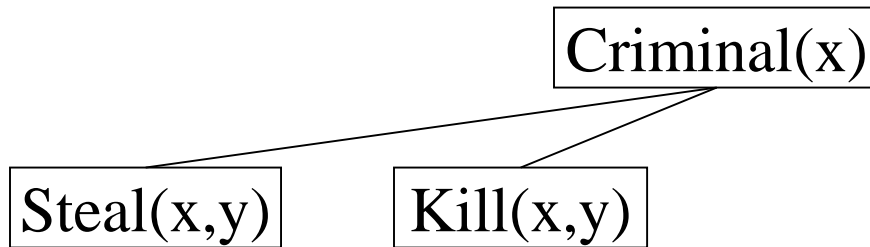
- Question: Has Reality Man done anything criminal?



FAIL

Backward Chaining

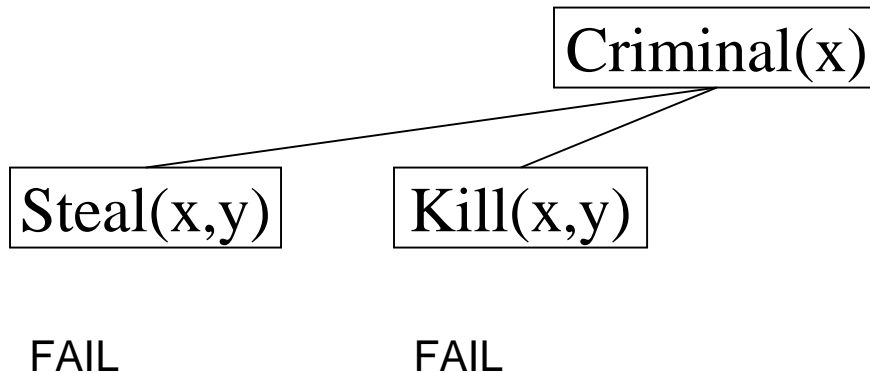
- Question: Has Reality Man done anything criminal?



FAIL

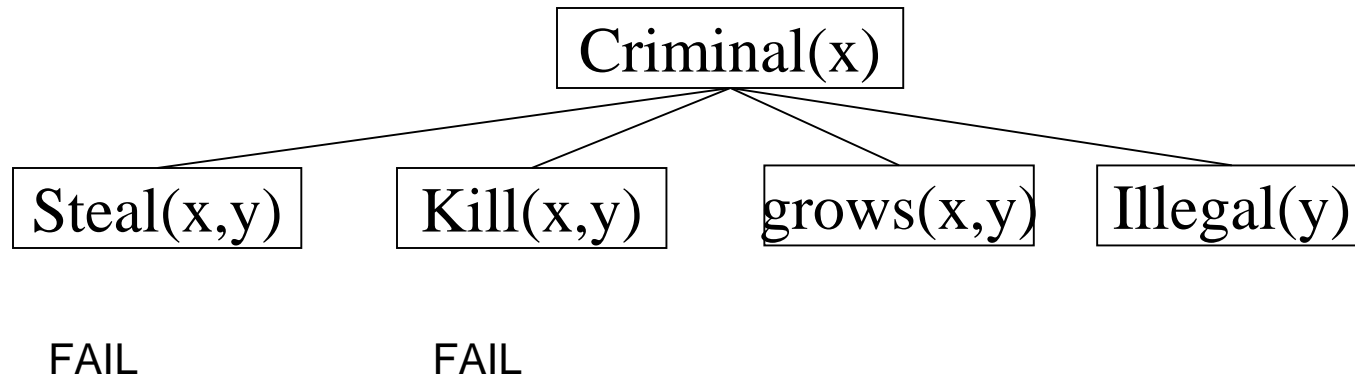
Backward Chaining

- Question: Has Reality Man done anything criminal?



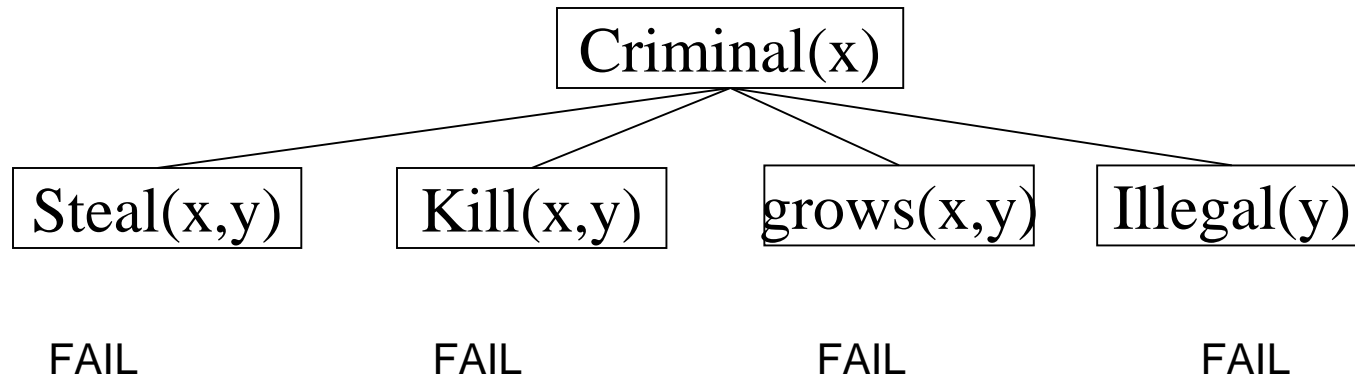
Backward Chaining

- Question: Has Reality Man done anything criminal?



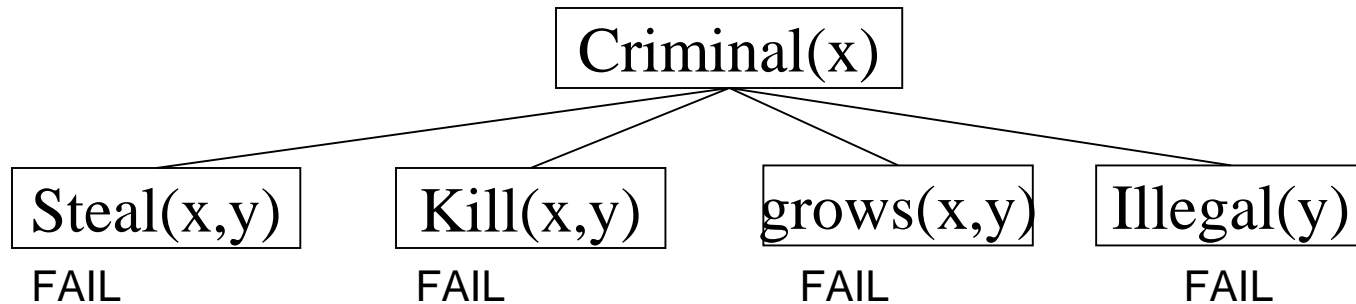
Backward Chaining

- Question: Has Reality Man done anything criminal?



Backward Chaining

- Question: Has Reality Man done anything criminal?



- Backward Chaining is a **depth-first search**: in any knowledge base of realistic size, many search paths will result in failure.

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 - **Resolution**

Resolution (a simple example)

KB:

(1) father (art, jon)

(2) father (bob, kim)

(3) father (X, Y) \Rightarrow parent (X, Y)

Goal: parent (art, jon)?

(KB) \wedge (\neg Goal) is "Unsatisfiable"

Resolution (a simple example)

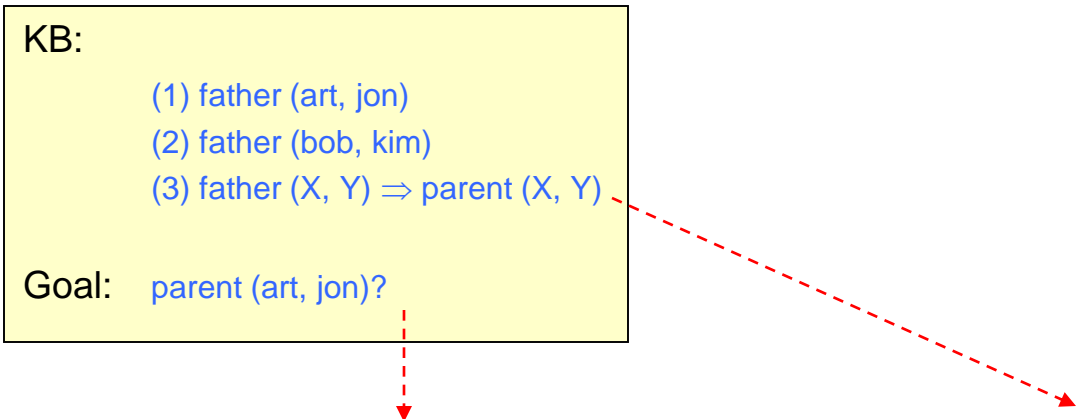
KB:

(1) father (art, jon)

(2) father (bob, kim)

(3) father (X, Y) \Rightarrow parent (X, Y)

Goal: parent (art, jon)?


$$\begin{array}{ccc} \neg \text{parent}(\text{art}, \text{jon}) & & \text{father}(\text{X}, \text{Y}) \Rightarrow \text{parent}(\text{X}, \text{Y}) \\ & \backslash \quad / & \\ \neg \text{father}(\text{art}, \text{jon}) & & \text{father}(\text{art}, \text{jon}) \\ & \backslash \quad / & \\ & [] & \end{array}$$

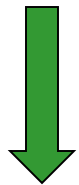
FOL resolution rule

$$\text{UNIFY}(\ell_i, \neg m_j) = \theta.$$

$$\frac{\ell_1 \vee \dots \vee \ell_k, \quad m_1 \vee \dots \vee m_n}{\text{SUBST}(\theta, \ell_1 \vee \dots \vee \ell_{i-1} \vee \ell_{i+1} \vee \dots \vee \ell_k \vee m_1 \vee \dots \vee m_{j-1} \vee m_{j+1} \vee \dots \vee m_n)}$$

- Example:

$$[\textit{Animal}(F(x)) \vee \textit{Loves}(G(x), x)] \quad \text{and} \quad [\neg \textit{Loves}(u, v) \vee \neg \textit{Kills}(u, v)]$$



$$\theta = \{u/G(x), v/x\}$$

$$[\textit{Animal}(F(x)) \vee \neg \textit{Kills}(G(x), x)]$$

FOL resolution (example)

$$\frac{\neg Rich(x) \vee Unhappy(x) \quad Rich(Me)}{Unhappy(Me)}$$

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

FOL Conjunctive normal form (CNF)

Steps:

1. Replace $P \Rightarrow Q$ by $\neg P \vee Q$
2. Move \neg inwards, e.g., $\neg \forall x P$ becomes $\exists x \neg P$
3. Standardize variables apart, e.g., $\forall x P \vee \exists x Q$ becomes $\forall x P \vee \exists y Q$
4. Move quantifiers left in order, e.g., $\forall x P \vee \exists x Q$ becomes $\forall x \exists y P \vee Q$
5. Eliminate \exists by Skolemization (next slide)
6. Drop universal quantifiers
7. Distribute \wedge over \vee , e.g., $(P \wedge Q) \vee R$ becomes $(P \vee Q) \wedge (P \vee R)$

Skolemization

- Why can't (y) be replaced by a constant symbol?

$$\forall x [\exists y \text{ Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists z \text{ Loves}(z, x)]$$

- Everyone loves the same animal (F), and the same (G) loves everyone

$$\forall x [\text{Animal}(F) \wedge \neg \text{Loves}(x, F)] \vee \text{Loves}(G, x)$$



- Each person loves a different animal $F(x)$, and a different $G(x)$ loves each person

$$\forall x [\text{Animal}(F(x)) \wedge \neg \text{Loves}(x, F(x))] \vee \text{Loves}(G(x), x)$$

- Distribution

$$\forall x [\text{Animal}(F(x)) \vee \text{Loves}(G(x), x)] \wedge [\neg \text{Loves}(x, F(x)) \vee \text{Loves}(G(x), x)]$$

Converting to CNF

$$\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$$

- Eliminate implications

$$\forall x [\neg \forall y \neg \text{Animal}(y) \vee \text{Loves}(x, y)] \vee [\exists y \text{ Loves}(y, x)]$$

- Move negation inwards

$$\forall x [\exists y \neg(\neg \text{Animal}(y) \vee \text{Loves}(x, y))] \vee [\exists y \text{ Loves}(y, x)] .$$

$$\forall x [\exists y \neg \neg \text{Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists y \text{ Loves}(y, x)] .$$

$$\forall x [\exists y \text{ Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists y \text{ Loves}(y, x)] .$$

- Standardize variables

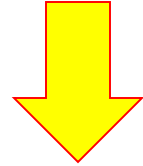
$$\forall x [\exists y \text{ Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists z \text{ Loves}(z, x)]$$

- Skolemization

$$\forall x [\text{Animal}(F(x)) \wedge \neg \text{Loves}(x, F(x))] \vee \text{Loves}(G(x), x)$$

FOL resolution (another example)

- A. $\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$
- B. $\forall x [\exists z \text{ Animal}(z) \wedge \text{Kills}(x, z)] \Rightarrow [\forall y \neg \text{Loves}(y, x)]$
- C. $\forall x \text{ Animal}(x) \Rightarrow \text{Loves}(\text{Jack}, x)$
- D. $\text{Kills}(\text{Jack}, \text{Tuna}) \vee \text{Kills}(\text{Curiosity}, \text{Tuna})$
- E. $\text{Cat}(\text{Tuna})$
- F. $\forall x \text{ Cat}(x) \Rightarrow \text{Animal}(x)$
- \neg G. $\neg \text{Kills}(\text{Curiosity}, \text{Tuna})$



Transform to CNF

FOL resolution (another example)

- A. $\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$
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- F. $\forall x \text{ Cat}(x) \Rightarrow \text{Animal}(x)$
- \neg G. $\neg \text{Kills}(\text{Curiosity}, \text{Tuna})$

- A1. $\text{Animal}(F(x)) \vee \text{Loves}(G(x), x)$
- A2. $\neg \text{Loves}(x, F(x)) \vee \text{Loves}(G(x), x)$
- B. $\neg \text{Loves}(y, x) \vee \neg \text{Animal}(z) \vee \neg \text{Kills}(x, z)$
- C. $\neg \text{Animal}(x) \vee \text{Loves}(\text{Jack}, x)$
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