

EECS 391: Introduction to AI

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Announcements

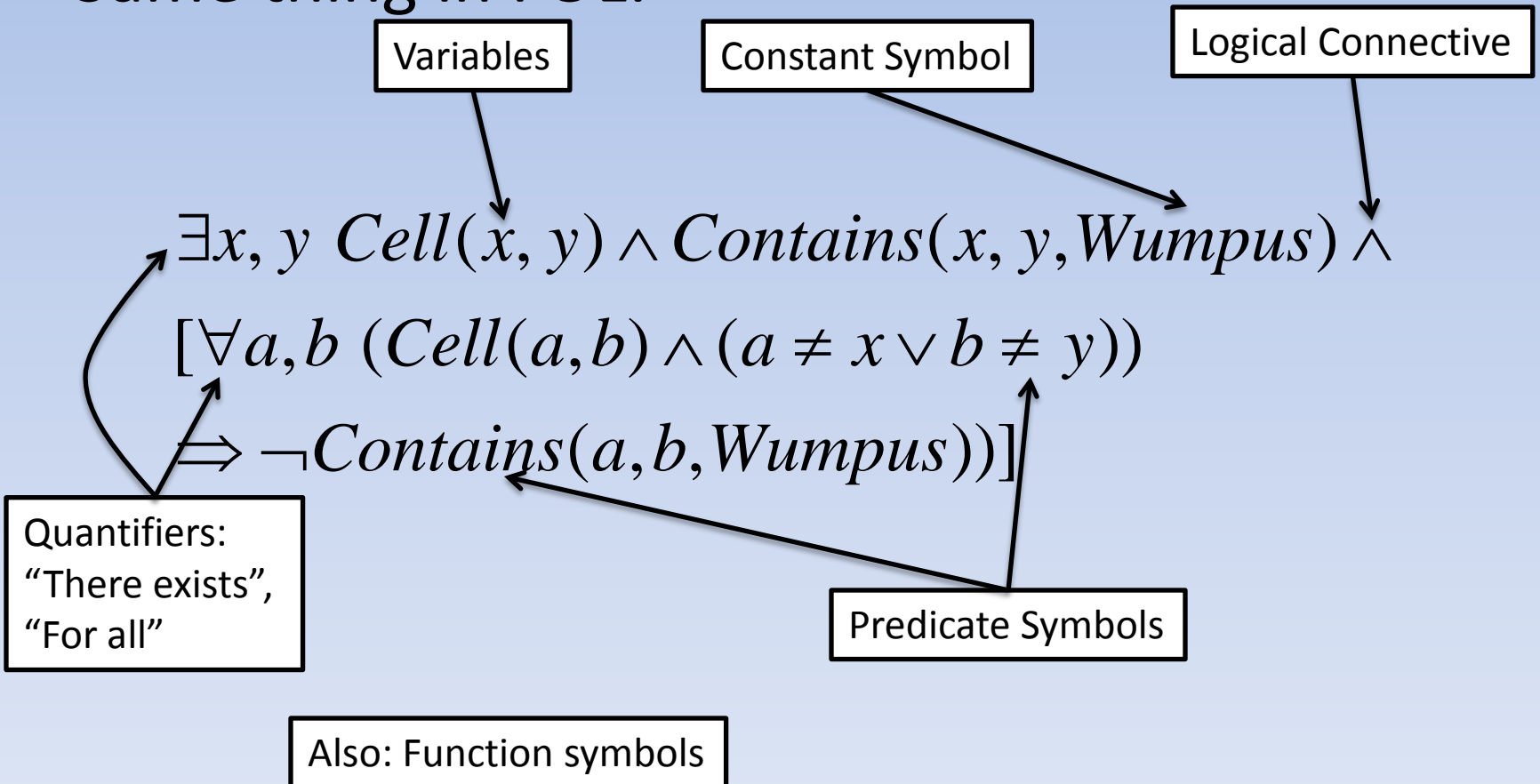
- HW3 due (solutions tomorrow noon)
- Quiz 3 Thursday

Today

- First order logic (Chapter 8)
- FOL Inference (Chapter 9)

First Order Logic

- Same thing in FOL:



Key Elements of FOL

- FOL introduces:
 - **Objects**
 - *Relations* between objects (predicates)
 - *Functions* that map objects to other objects
 - *Variables* that act as placeholders for objects
 - *Quantifiers* that allow us to talk about (infinite) collections of objects

Functions and Predicates

- A **function** takes a set of objects and returns an object, e.g. $Mother(A)=B$
- A **predicate** takes a set of objects and returns true/false, e.g. $isMotherOf(A,B)?$

Syntax

- Basic elements:
 - Constant symbols (representing objects in the world)
 - Predicate symbols (representing relations between objects)
 - Function symbols (can be used to map objects to others)
- Also variables (placeholders for objects)

Terms

- A constant or variable symbol is a term
- If t_1, \dots, t_n are terms and f is a function of arity n , $f(t_1, \dots, t_n)$ is a term
- Examples: *Wumpus*, *Pit*, x , y , *XWestOf*(x, y)
- A term is a generic representation of an object

Atomic Formulae

- If P is a predicate symbol of arity n , and t_1, \dots, t_n are terms, $P(t_1, \dots, t_n)$ is an atomic formula
- Examples: *Contains*($x, y, Wumpus$),
Smelly(x, y), *Breezy*(x, y)
- Analogous to proposition symbols in propositional logic

Complex Formulae

- All atomic formulae are complex formulae
- If F_1 and F_2 are formulae, so is $(\neg F_1)$, $(F_1 \wedge F_2)$, etc (for all connectives)
- If F_1 is a formula, so is $(\forall x F_1)$ and $(\exists x F_1)$, where x is a variable symbol
- Example:

$$\forall x \forall y (Cell(x, y) \wedge Contains(x, y, Pit) \wedge Cell(XNorthOf(x, y), YNorthOf(x, y))) \Rightarrow Breezy(XNorthOf(x, y), YNorthOf(x, y))$$

Quantifiers

- FOL has “quantifiers” \forall and \exists
- \forall is the “universal quantifier”, read “For all”
 - $\forall xP(x)$ is a formula that says P is true for every object x
 - Can be thought of as a conjunction over the universe of objects: $P(a) \wedge P(b) \wedge P(c)\dots$
- \exists is the “existential quantifier”, read “There exists”
 - $\exists xP(x)$ is a formula that says there is some object x for which P is true
 - Can be thought of as a disjunction over the universe of objects: $P(a) \vee P(b) \vee P(c)\dots$

Quantifier nesting

- Quantifiers can be nested:

$\exists x \exists y \textit{Contains}(x, y, \textit{Wumpus})$

(or $\exists x, y \textit{Contains}(x, y, \textit{Wumpus})$)

- The order of nesting is important
 - $\forall x \exists y \textit{Friend}(x, y)$: “Everybody has a friend”
 - $\exists y \forall x \textit{Friend}(x, y)$: “There is someone who is everyone’s friend”

Quantifier Relationships

- Observe: if “for all x , $P(x)$ ”, then there is no x for which $\neg P(x)$
- So $\forall x P(x)$ is the same as $\neg \exists y \neg P(y)$
- Similarly, $\exists x P(x)$ is the same as $\neg \forall y \neg P(y)$
- However, using the two symbols improves readability, so we’ll keep them

Variable scoping

- If two quantifiers use the same variable, the variable “belongs” to the innermost quantifier

$$\begin{aligned} \forall x P(x) \wedge \exists x \forall y Q(x, y) &\equiv \\ \forall x P(x) \wedge \exists a \forall y Q(a, y) & \end{aligned}$$

- Generally bad idea, don't do this
 - Variables are placeholders, so can always use new variables as needed
 - Only use the same variable if you need to talk about the same object

Terminology

- If a formula has no variables, it is said to be a “ground” formula (vs non-ground) (also “ground fact” or “fact”)
 - Contains(1,2,Wumpus) vs Contains(x, y, Wumpus)
- A variable that appears within the scope of some quantifier is “bound” (vs free)
 - $\exists x, y$ Contains(x, y, Wumpus) vs. Contains(x, y, Wumpus)
- A formula with only bound variables or constants is said to be “closed”
 - We will always work with closed formulae

Semantics of FOL

- In propositional logic, a model assigned truth values to propositional symbols
- FOL is much more expressive; a model contains objects that the constant symbols refer to (could be infinite!)
 - The “domain” of a model is the set of objects it has
 - Example: In the wumpus world, the objects are the wumpus, pits, and the numbers representing grid coordinates

Predicates and functions

- A predicate $P(t_1, \dots, t_n)$ is a list of tuples of objects for which the relation holds
- A function $f(t_1, \dots, t_n)$ is a mapping from the (cross product of) model's domain to itself
- Together with the assignments for constant symbols, these constitute an “interpretation”

Examples

- Contains (x, y, z)

X	Y	Z
3	2	Wumpus
1	3	Pit
(etc)		

- XWestOf(x, y)

x	y	XWestOf(X,Y,Wumpus)
1	1	-1
2	2	1
3	1	2
(etc)		

Compositional Semantics

- FOL semantics are compositional
 - To find the object for $f(t_1, \dots, t_n)$, find the objects for t_1, \dots, t_n
 - Then look up the table for f to find the object for $f(t_1, \dots, t_n)$
 - Likewise for predicates

Semantics of a formula

- For atomic formula $P(t_1, \dots, t_n)$: Given the interpretation of the terms, does the tuple appear in the interpretation's list for P ?
- For complex formulae:
 - Use semantics for logical connectives
 - Universal Quantifier: Formula holds if it is true for all possible **bindings** in the given interpretation
 - Existential Quantifier: Formula holds if a binding exists that makes it true in the given interpretation

Higher order logic

- Propositional logic (“zeroth-order logic”) has no concept of objects
- FOL allows objects
- Second order logic allows relations over objects to be objects
 - E.g., can say “there exists a relation that is transitive”

FOL Inference

- As before, we want to find out if certain formulae are entailed by what we know
- In propositional logic, all we were interested in was whether a formula was entailed
- In FOL, we are generally also interested in *what variable bindings lead to entailment*

FOL Inference

$\forall x \textit{Feathers}(x) \wedge \textit{Flies}(x) \Rightarrow \textit{Bird}(x)$

$\textit{Feathers}(\textit{Crow})$

$\textit{Flies}(\textit{Crow})$

$\textit{Feathers}(\textit{Ostrich})$

$? \exists x \textit{Bird}(x)$

FOL Inference Roadmap

- If we can somehow turn our KB into a propositional KB, we can use techniques we already know—*propositionalization*
 - This has some drawbacks
- Then we'll see how to extend resolution to FOL

Converting FOL to propositions

- Dealing with \forall :
 - If our KB has $\forall xP(x)$, we can *substitute* x with all objects in our model's domain
 - replace it with a sequence of facts: $P(a)$, $P(b)$, $P(c)$, ...
 - Notice this could be infinite if there are function symbols: $P(a)$, $P(f(a))$, $P(f(f(a)))$, ...

Substitution

- We write $p\{v/c\}$ to mean the formula p with the variable v substituted with the term c
 - *Simultaneously* replace every occurrence of v with c
 - c is a **binding** for v
- Examples: Suppose $\theta = \{a/f(b,c), b/c\}$,
 $\sigma = \{b/Const\}$
 - Then $p(a,b)\theta = p(f(b,c),c)$, $p(a,b)\sigma = p(a,Const)$

Converting FOL to propositions

- Dealing with \exists :
 - If our KB has $\exists xP(x)$, we can *substitute* x with *some* object from our model's domain
 - We don't specifically know the name of this object, so let's invent one
 - This is called a "Skolem constant"
 - So $\exists xP(x)$ becomes $P(\text{NewSymbol})$, where *NewSymbol* is a constant symbol not used by anything else in our KB

Converting FOL to propositions

- At this point, we've eliminated all quantifiers and are left with a collection of *ground* formulae
- We can treat each atomic formula as a separate propositional symbol
 - This reduces the KB to a propositional one
 - Then to answer questions we can use propositional inference

Example

$\forall x \text{ Feathers}(x) \wedge \text{Flies}(x) \Rightarrow \text{Bird}(x)$

$\text{Feathers}(\text{Crow})$

$\text{Flies}(\text{Crow})$

$\text{Feathers}(\text{Ostrich})$

$\text{Feathers}(\text{Crow})$

$\text{Flies}(\text{Crow})$

$\text{Feathers}(\text{Ostrich})$

$\text{Feathers}(\text{Crow}) \wedge \text{Flies}(\text{Crow}) \Rightarrow \text{Bird}(\text{Crow})$

$\text{Feathers}(\text{Ostrich}) \wedge \text{Flies}(\text{Ostrich}) \Rightarrow \text{Bird}(\text{Ostrich})$

Example

Feathers _ Crow

Flies _ Crow

Feathers _ Ostrich

Feathers _ Crow \wedge *Flies _ Crow* \Rightarrow *Bird _ Crow*

Feathers _ Ostrich \wedge *Flies _ Ostrich* \Rightarrow *Bird _ Ostrich*

Issues

- What if there are function symbols?
 - There are theorems (Skolem, Herbrand, Godel 1929-1930) that says that if a FOL formula is entailed by a KB, it must be provable using just a finite subset of the propositionalized KB
 - We could do a sort of “iterative deepening” of the KB
- So if a formula is entailed, it can be proved eventually---completeness!
- But what if it is *not* entailed?

Entailment is Semidecidable

- If a formula is *not* entailed by a KB, in general there is no way to tell by an algorithm
 - i.e., no algorithm exists that will always finitely determine non-entailment
 - Our “iterative deepening” procedure might in general run forever
- This creates problems for real logic systems
 - Prolog has “negation as failure”

Issues Continued

- Converting to a propositionalized KB is very expensive
- Results in huge propositional KBs even for small first order KBs
 - Contains large numbers of formulae that might be completely irrelevant to query
 - Still sometimes usable due to fast propositional SAT solvers
- These issues can be solved by FOL (syntactic) inference algorithms

Lifting

- To develop syntactic FOL inference, we can take the inference rules we know about from propositional logic and make FOL versions of them
 - This is called “**lifting**”

(Lifted) Resolution

- Like in propositional logic, resolution is a general purpose proof procedure
- Sound and refutation complete
- In general, not efficient; however, special versions of resolutions exist that trade off completeness and efficiency

Resolution inference rule

$$\frac{l_1 \vee l_2 \vee m \dots \vee l_k, \quad r_1 \vee r_2 \vee \neg m \dots \vee r_k}{l_1 \vee l_2 \dots \vee l_k \vee r_1 \vee r_2 \dots \vee r_k}$$

$$\frac{l_1 \vee l_2 \vee m_i \dots \vee l_k, \quad r_1 \vee r_2 \vee \neg n_j \dots \vee r_k, \quad m_i \theta = \neg n_j \theta}{l_1 \vee l_2 \dots \vee l_k \vee r_1 \vee r_2 \dots \vee r_k \{ \theta \}}$$

Atomic Formulae

Unification: This substitution makes m_i identical to $\neg n_j$