

CMSC 471: Intro to Al

First-Order Logic



FOL Overview

- First Order logic (FOL) is a powerful knowledge representation (KR) system
- Used in AI systems in various ways, e.g., to
- -Directly represent & reason about concepts & objects
- Formally specify meaning of KR systems (e.g., <u>OWL</u>)
- -Form programming languages (e.g., Prolog) and rule-based systems
- Make semantic database systems (<u>Datalog</u>) and Knowledge graphs (<u>Wikidata</u>)
- Provide features useful in neural network deep learning systems



First-order logic

- First-order logic (FOL) models the world in terms of
 - Objects, which are things with individual identities
 - Properties of objects that distinguish them from others
 - Relations that hold among sets of objects
 - Functions, a subset of relations where there is only one "value" for any given "input"
- Examples:
 - Objects: students, lectures, companies, cars ...
 - Relations: isa, hasBrother, biggerThan, outside, hasPart, color, occursAfter, owns, visits, precedes, ...
 - Properties: blue, oval, even, large, ...
 - Functions: hasFather, hasSSN, ...



User provides

- Constant symbols representing individuals in world
 - BarackObama, Green, John, 3, "John Smith"
- Predicate symbols map individuals to truth values
 - greater(5,3)
 - green(Grass)
 - color(Grass, Green)
 - hasBrother(John, Robert)
- Function symbols map individuals to individuals
 - hasFather(SashaObama) = BarackObama
 - colorOf(Sky) = Blue



What do these mean?

- User should also indicate what these mean in a way that humans will understand
 - i.e., map to their own internal representations
- May be done via a combination of
 - Choosing good names for formal terms, e.g. calling a concept
 HumanBeing instead of Q5
 - Comments in the definition #human being
 - Descriptions and examples in documentation
 - Reference to other representations , e.g., sameAs /m/0dgw95
 in Freebase and Person in schema.org
 - Give examples like *Donald Trump* and *Luke Skywalker* to help distinguish the concepts of a real and fictional person

FOL Provides

- Variable symbols
 - e.g., X, Y, ?foo, ?number
- Connectives
 - Same as propositional logic: not (¬), and (∧), or
 (∨), implies (→), iff (↔), equivalence (≡), ...
- Quantifiers
 - Universal $\forall x$ or (Ax)
 - Existential ∃x or (Ex)



Sentences: built from terms and atoms

- **term** (denoting an individual): constant or variable symbol, or n-place function of n terms, e.g.:
 - Constants: john, umbc
 - Variables: X, Y, Z
 - Functions: mother_of(john), phone(mother(x))
- Ground terms have no variables in them
 - Ground: john, father_of(father_of(john))
 - Not Ground: father_of(X)
- Syntax may vary: maybe variables must start with a "?" or a capital letter

Sentences: built from terms and atoms

- atomic sentences (which are either true or false) are nplace predicates of n terms, e.g.:
 - green(kermit)
 - between(philadelphia, baltimore, dc)
 - loves(X, mother(X))
- complex sentences formed from atomic ones connected by the standard logical connectives with quantifiers if there are variables, e.g.:
 - loves(mary, john) ∨ loves(mary, bill)
 - $\forall x loves(mary, x)$



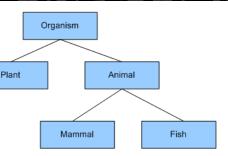
What do atomic sentences mean?

- Unary predicates typically encode a type
 - muppet(Kermit): kermit is a kind of muppet
 - green(kermit): kermit is a kind of green thing
 - integer(X): x is a kind of integer
- Non-unary predicates typically encode relations or properties
 - Loves(john, mary)
 - Greater_than(2, 1)
 - Between(newYork, philadelphia, baltimore)
 - hasName(john, "John Smith")



Ontology

- Designing a logic representation is like designing a model in an object-oriented language
- Ontology: a "formal naming and definition of the types, properties and relations of entities for a domain of discourse"
- E.g.: <u>schema.org</u> ontology used to put semantic data on Web pages to help search engines
 - Here's the <u>semantic markup</u> Google sees about <u>CSEE</u> <u>department of UMBC</u>.





Sentences: built from terms and atoms

quantified sentences adds quantifiers ∀ and ∃
 ∀x loves(x, mother(x))

 $\exists x \text{ number}(x) \land \text{greater}(x, 100), \text{prime}(x)$

well-formed formula (wff): a sentence with no free variables or where all variables are bound by a universal or existential quantifier
 In (∀x)P(x, y) x is bound & y is free so it's not a wff



Quantifiers: \forall and \exists

Universal quantification

- $(\forall x)P(X)$ means P holds for **all** values of X in the domain associated with variable¹
- E.g., $(\forall X)$ dolphin $(X) \rightarrow$ mammal(X)

Existential quantification

- (∃x)P(X) means P holds for **some** value of X in domain associated with variable
- E.g., (∃**X**) mammal(X) \land lays_eggs(X)
- This lets us make statements about an object without identifying it

¹ a variable's domain is often not explicitly stated and is assumed by the context



Universal Quantifier: ∀

Universal quantifiers typically used with implies to form rules:

Logic: $\forall X \text{ student}(X) \rightarrow \text{smart}(X)$

Means: All students are smart

Universal quantification rarely used without implies:

Logic: $\forall X$ student(X) \land smart(X)

Means: Everything is a student and is smart

What about this, though:

-Logic: \forall X alive(X) ∨ dead(X)

–Means: everything is either alive or dead

Universal Quantifier: ∀

- What about this, though:
 - -Logic: ∀X alive(X) ∨ dead(X)
 - -Means: everything is either alive or dead
- Can be rewritten using a standard tautology

$$-A \vee B \equiv ^{\sim}A \rightarrow B$$

- Giving both of these (since $A \lor B \equiv B \lor A$)
 - $\forall X \sim alive(X) \rightarrow dead(X)$
 - $\forall X \sim dead(X) \rightarrow alive(X)$



Existential Quantifier: ∃

 Existential quantifiers usually used with and to specify a list of properties about an individual

Logic: $(\exists X)$ student(X) \land smart(X)

Meaning: There is a student who is smart

• Common mistake: represent this in FOL as:

Logic: $(\exists X)$ student $(X) \rightarrow smart(X)$

Meaning: ?

Existential Quantifier: ∃

 Existential quantifiers usually used with and to specify a list of properties about an individual

Logic: $(\exists X)$ student(X) \land smart(X)

Meaning: There is a student who is smart

Common mistake: represent this in FOL as:

Logic: $(\exists X)$ student $(X) \rightarrow smart(X)$

$$P \rightarrow Q = {}^{\sim}P \vee Q$$

 $\exists X \ student(X) \rightarrow smart(X) = \exists X \ \sim student(X) \ v \ smart(X)$

Meaning: There's something that is either not a student or is smart

Quantifier Scope

- FOL sentences have structure, like programs
- In particular, variables in a sentence have a scope
- Suppose we want to say "everyone who is alive loves someone"
 (∀X) alive(X) → (∃ Y) loves(X, Y)
- Here's how we scope the variables

 $(\forall X) \text{ alive}(X) \rightarrow (\exists Y) \text{ loves}(X, Y)$





Quantifier Scope

- Switching order of two universal quantifiers does not change the meaning
 - $(\forall X)(\forall Y)P(X,Y) \leftrightarrow (\forall Y)(\forall X) P(X,Y)$
 - Dogs hate cats (i.e., all dogs hate all cats)
- You can switch order of existential quantifiers
 - $(\exists X)(\exists Y)P(X,Y) \leftrightarrow (\exists Y)(\exists X) P(X,Y)$
 - A cat killed a dog
- Switching order of universal and existential quantifiers does change meaning:
 - Everyone likes someone: $(\forall X)(\exists Y)$ likes(X,Y)
 - Someone is liked by everyone: (∃Y)(\forall X) likes(X,Y)



```
def verify1():
# Everyone likes someone: (\forall x)(\exists y) likes(x,y)
for p1 in people():
  foundLike = False
                                Every person has at
  for p2 in people():
                                least one individual that
     if likes(p1, p2):
                                they like.
       foundlike = True
       break
  if not foundlike:
     print(p1, 'does not like anyone ⊗')
     return False
return True
```

Procedural example 1



```
def verify2():
# Someone is liked by everyone: (\exists y)(\forall x) likes(x,y)
for p2 in people():
  foundHater = False
                                      There is a person who is
  for p1 in people():
                                      liked by every person in
     if not likes(p1, p2):
                                      the universe.
       foundHater = True
       break
  if not foundHater
     print(p2, 'is liked by everyone ©')
     return True
return False
```

Procedural example 2

Connections between \forall and \exists

- We can relate sentences involving ∀ and ∃ using extensions to De Morgan's laws:
 - 1. $(\forall x) P(x) \leftrightarrow \neg(\exists x) \neg P(x)$
 - 2. $\neg(\forall x) P(x) \leftrightarrow (\exists x) \neg P(x)$
 - 3. $(\exists x) P(x) \leftrightarrow \neg (\forall x) \neg P(x)$
 - 4. $\neg(\exists x) P(x) \leftrightarrow (\forall x) \neg P(x)$
- Examples
 - 1. All dogs sleep ↔ There is no dog that doesn't sleep
 - 2. Not all dogs bark ↔ There is a dog that doesn't bark
 - 3. There is a dog that talks ↔ Not all dogs can't talk
 - 4. No dog likes cats ↔ All dogs don't like cats



Every gardener likes the sun

All purple mushrooms are poisonous



Every gardener likes the sun

 $\forall x \text{ gardener}(x) \rightarrow \text{likes}(x,Sun)$

All purple mushrooms are poisonous



Every gardener likes the sun

 $\forall x \text{ gardener}(x) \rightarrow \text{likes}(x,\text{Sun})$

All purple mushrooms are poisonous

 $\forall x (mushroom(x) \land purple(x)) \rightarrow poisonous(x)$



Every gardener likes the sun

 $\forall x \text{ gardener}(x) \rightarrow \text{likes}(x,\text{Sun})$

All purple mushrooms are poisonous

 $\forall x (mushroom(x) \land purple(x)) \rightarrow poisonous(x)$

No purple mushroom is poisonous (two ways)



Every gardener likes the sun

 $\forall x \text{ gardener}(x) \rightarrow \text{likes}(x,\text{Sun})$

All purple mushrooms are poisonous

 $\forall x (mushroom(x) \land purple(x)) \rightarrow poisonous(x)$

No purple mushroom is poisonous (two ways)

 $\neg \exists x \text{ purple}(x) \land \text{mushroom}(x) \land \text{poisonous}(x)$



Every gardener likes the sun

 $\forall x \text{ gardener}(x) \rightarrow \text{likes}(x,Sun)$

All purple mushrooms are poisonous

 $\forall x (mushroom(x) \land purple(x)) \rightarrow poisonous(x)$

No purple mushroom is poisonous (two ways)

 $\neg \exists x \text{ purple}(x) \land \text{mushroom}(x) \land \text{poisonous}(x)$

 $\forall x \ (mushroom(x) \land purple(x)) \rightarrow \neg poisonous(x)$



English to FOL: Counting

Use = predicate to identify different individuals



- There are <u>at least</u> two purple mushrooms
 - $\exists x \exists y \text{ mushroom}(x) \land \text{purple}(x) \land \text{mushroom}(y) \land \text{purple}(y) \land \neg (x=y)$
- There are <u>exactly</u> two purple mushrooms

```
\exists x \exists y \; mushroom(x) \land purple(x) \land mushroom(y) \land purple(y) \land \neg(x=y) \land \forall z \; (mushroom(z) \land purple(z)) \rightarrow ((x=z) \lor (y=z))
```

Saying there are 802 different <u>Pokemon</u> is hard! Direct use of FOL is not for everything!



What do these mean?



You can fool some of the people all of the time

You can fool all of the people some of the time



What do these mean?

Both English statements are ambiguous



There is a nonempty subset of people so easily fooled that you can fool that subset every time.

For any given time, there is a non-empty subset at that time that you can fool

You can fool all of the people some of the time

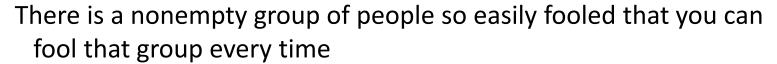
There are one or more times when it's possible to fool everyone.

Each individual can be fooled at some point in time





You can fool some of the people all of the time



≡ There's (at least) one person you can fool every time

 $\exists x \ \forall t \ person(x) \land time(t) \rightarrow canFool(x, t)$

For any given time, there is a non-empty group at that time that you can fool

≡ For every time, there's a person at that time that you can fool

 $\forall t \exists x \ person(x) \land time(t) \rightarrow canFool(x, t)$





You can fool *all of* the people *some of* the time

There's at least one time when you can fool everyone

 $\exists t \ \forall x \ time(t) \land person(x) \rightarrow canFool(x, t)$

Everybody can be fooled at some point in time

 $\forall x \exists t \ person(x) \land time(t) \rightarrow canFool(x, t)$



Representation Design

Many options for representing even a simple fact, e.g., something's color as red, green or blue, e.g.:



- green(kermit)
- color(kermit, green)
- hasProperty(kermit, color, green)
- Choice can influence how easy it is to use
- Last option of representing properties & relations as <u>triples</u> used by modern <u>knowledge graphs</u>
 - Easy to ask: What color is Kermit? What are Kermit's properties?,
 What green things are there? Tell me everything you know, ...



Simple genealogy KB in FOL

Design a knowledge base using FOL that

- Has facts of immediate family relations, e.g., spouses, parents, etc.
- Defines more complex relations (ancestors, relatives)
- Detect conflicts, e.g., you are your own parent
- Infers relations, e.g., grandparent from parent
- Answers queries about relationships between people



Extend with relations and constraints

- Simple two argument relations, e.g.
 - spouse, has_child, has_parent
- Add general constraints to relations, e.g.
 - spouse(X,Y) => \sim (X = Y)
 - spouse(X,Y) => person(X) \land person(Y)
- Add FOL sentences for inference, e.g.
 - spouse(X,Y) \Leftrightarrow spouse(Y,X)
- Add instance data
 - e.g., spouse(Djt, Mt)



Example: A simple genealogy KB in FOL

Predicates:

- -parent(x, y), child(x, y), father(x, y), daughter(x, y), etc.
- -spouse(x, y), husband(x, y), wife(x,y)
- -ancestor(x, y), descendant(x, y)
- -male(x), female(y)
- -relative(x, y)

Facts (ground terms):

- —husband(Joe, Mary), son(Fred, Joe)
- -spouse(John, Nancy), male(John), son(Mark, Nancy)
- —father(Jack, Nancy), daughter(Linda, Jack)
- -daughter(Liz, Linda)
- -etc.



Example Axioms

```
(\forall x,y) parent(x, y) \leftrightarrow child (y, x)
```

```
(\forall x,y) father(x, y) \leftrightarrow parent(x, y) \land male(x)
```

$$(\forall x,y)$$
 mother $(x,y) \leftrightarrow parent(x,y) \land female(x)$

$$(\forall x,y)$$
 daughter $(x,y) \leftrightarrow \text{child}(x,y) \land \text{female}(x)$

$$(\forall x,y) \text{ son}(x,y) \leftrightarrow \text{child}(x,y) \land \text{male}(x)$$

$$(\forall x,y)$$
 husband $(x,y) \leftrightarrow \text{spouse}(x,y) \land \text{male}(x)$

$$(\forall x,y)$$
 spouse $(x, y) \leftrightarrow$ spouse (y, x)

...



Axioms, definitions and theorems

- Axioms: facts and rules that capture (important) facts & concepts in a domain; used to prove theorems
- Dependent axioms can make reasoning faster, however
- Choosing a good set of axioms is a 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 "p(x) ← ..."
 - Some concepts have definitions (e.g., triangle) and some don't (e.g., person)

More on definitions

Example: define father(x, y) by parent(x, y) & male(x)

- parent(x, y) is a necessary (but not sufficient) description of father(x, y)
 father(x, y) → 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) ^{\wedge} male(x) \leftrightarrow father(x, y)
```

Higher-order logic

- FOL only lets us quantify over variables, and variables can only range over objects
- HOL allows us to quantify over relations, e.g.
 - "two functions are equal iff they produce the same value for all arguments"

$$\forall f \ \forall g \ (f = g) \leftrightarrow (\forall x \ f(x) = g(x))$$

• E.g.: (quantify over predicates)

```
\forallr transitive(r) \rightarrow (\forallxyz) r(x,y) \wedge r(y,z) \rightarrow r(x,z))
```

More expressive, but reasoning is undecide-able, in general



Examples of FOL in use



- Semantics of W3C's <u>Semantic Web</u> stack (RDF, RDFS, OWL) is defined in FOL
- OWL Full is equivalent to FOL
- Other OWL profiles support a subset of FOL and are more efficient
- FOL oriented knowledge representation systems have many user friendly tools
- E.g.: Protégé for creating, editing and exploring OWL ontologies



Examples of FOL in use

Many practical approaches embrace the approach that "some data is better than none"

- The semantics of <u>schema.org</u> is only defined in natural language text
- Wikidata's knowledge graph has a rich schema
 - Many constraint/logical violations are flagged with warnings
 - However, not all, see this Wikidata query that finds people who are their own mother or father





Automated inference for FOL

- Automated inference for FOL is harder than PL
 - Variables can take on an infinite number of possible values from their domains
 - Hence there are potentially an infinite number of ways to apply the Universal Elimination rule
- Godel's Completeness Theorem says that FOL entailment is only semi-decidable
 - If a sentence is true given a set of axioms, there is a procedure that will determine this
 - If a sentence is false, there's no guarantee a procedure will ever discover this — it may never halt

Generalized Modus Ponens (GMP)

- Modus Ponens: P, P=>Q |= Q
- Generalized Modus Ponens extends this to rules in FOL
- Combines And-Introduction, Universal-Elimination, and Modus Ponens, e.g.
 - given P(c), Q(c), $\forall x P(x) \land Q(x) \rightarrow R(x)$
 - derive R(c)
- Must deal with
 - more than one condition on rule's left side
 - variables



Forward & Backward Reasoning

- We often talk about two reasoning strategies:
 - Forward chaining and
 - Backward chaining
- Both are equally powerful, but optimized for different use cases
- You can also have a mixed strategy



Forward chaining



- Proofs start with given axioms/premises in KB, deriving new sentences using GMP until the goal/query sentence is derived
 - The process follows a chain of rules and facts going from the KB to the conclusion
- This defines a forward-chaining inference procedure because it moves "forward" from the KB to the goal [eventually]
- Inference using GMP is sound and complete for KBs containing only Horn clauses

Forward chaining example

- KB:
 - allergies(X) \rightarrow sneeze(X)
 - $cat(Y) \land allergicToCats(X) \rightarrow allergies(X)$
 - cat(felix)
 - allergicToCats(mary)
- Goal:
 - sneeze(mary)



Backward chaining



- Backward-chaining deduction using GMP is also complete for KBs containing only Horn clauses
- Proofs start with the goal query, find rules with that conclusion, and then tries to prove each of the antecedents in the rule
- Keep going until you reach premises
- Avoid loops by checking if new subgoal is already on the goal stack
- Avoid repeated work: use a cache to check if new subgoal already proved true or failed

Backward chaining example

- KB:
 - allergies(X) \rightarrow sneeze(X)
 - $cat(Y) \land allergicToCats(X) \rightarrow allergies(X)$
 - cat(felix)
 - allergicToCats(mary)
- Goal:
 - sneeze(mary)



Forward vs. backward chaining

- Forward chaining is data-driven
 - Automatic, unconscious processing, e.g., object recognition, routine decisions
 - -May do lots of work that is irrelevant to the goal
 - -Efficient when you want to compute all conclusions
- Backward chaining is goal-driven, better for problemsolving and query answering
 - -Where are my keys? How do I get to my next class?
 - -Complexity can be much less than linear w.r.t KB size
 - -Efficient when you want one or a few decisions
 - -Good where the underlying facts are changing



Mixed strategy

- Many practical reasoning systems do both forward and backward chaining
- The way you encode a rule determines how it is used, as in % this is a forward chaining rule spouse(X,Y) => spouse(Y,X).
 % this is a backward chaining rule wife(X,Y) <= spouse(X,Y), female(X).
- Given a model of the rules you have and the kind of reason you need to do, it's possible to decide which to encode as FC and which as BC rules.



Summary

- Logical agents apply inference to KB 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 produce all entailed sentences
- FC and BC linear time, complete for Horn clauses
- Resolution is sound and complete for propositional and first-order logic