Introduction to Artificial Intelligence

Lecture 9 – Logical reasoning

CS/CNS/EE 154

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First order logic (FOL)

- Propositional logic is about simple facts
 - "There is a breeze at location [1,2]"
- First order logic is about facts involving
 - Objects: Numbers, people, locations, time instants, ...
 - Relations: Alive, IsNextTo, Before, ...
 - Functions: MotherOf, BestFriend, SquareRoot, OneMoreThan,
 ...
- Will be able to say:
 - IsBreeze(x); IsPit(x); IsNextTo(x,y)

$$\forall x, y : (IsPit(x) \land IsNextTo(x, y)) \Rightarrow IsBreeze(y)$$

FOL: Basic syntactic elements

Constants: KingJohn, 1, 2, ..., [1,1], [1,2], ...,[n,n], ...

Variables:
x, y, z, ...

• Predicates: Brother, >, =, ...

Functions: LeftLegOf, MotherOf, Sqrt, ...

■ Connectives: ∧, ∨, ¬

• Quantifiers: \forall, \exists

 Constant, predicates and functions are mere symbols (i.e., have no meaning on their own)

FOL Syntax: Atomic sentences

A (variable-free) term is a

- constant symbol or
- k-ary function symbol: function(term₁, term₂, ..., term_k)

Example: LeftLegOf(KingJohn)

An atomic sentence is a predicate symbol applied to terms

Example:

- Brother(KingJohn, RichardLionheart)
- IsNextTo([1,1],[1,2])
- > (Length(LeftLegOf(KingJohn)), Length(LeftLegOf (RichardLionheart)))

Models in FOL

- Much more complicated than in Propositional Logic
- Models contain
 - Set of objects (finite or countable)
 - Set of relations between objects (map obj's to truth values)
 - Set of functions (map objects to other objects)

and their interpretations:

- Mapping from constant symbols to model objects
- Mapping from predicate symbols to model relations
- Mapping from function symbols to model functions
- An atomic sentence predicate(term₁, term₂, ..., term_k) is true if the objects referred to by term₁, term₂, ..., term_k are in the relation referred to by predicate

Quantifiers

Allow variables in addition to constants

- Sentences with free variables: S(x)
- Quantifiers bind free variables

 $\forall x: S(x)$ is true if S(x) is true for all instantiations of x (i.e., for each possible object in the model)

 $\exists x: S(x)$ is true if S(x) is true for at least one instantiation of x (i.e., for some object)

- Example:
 - All homeworks in 154 are hard
 ∀x: (Homework (K, (54) >> Hand(K))
 - At least one of the 154 homeworks is hard

 At least one of the 154 homework is hard

 At least one of the 154 homework (x, 154) 1 Hard (x)

Properties of quantifiers

- Is $\forall x \ \forall y \ S(x,y)$ the same as $\forall y \ \forall x \ S(x,y)$?
- Is $\exists x \ \exists y \ S(x,y)$ the same as $\exists y \ \exists x \ S(x,y)$?
- Is $\exists x \ \forall y \ S(x,y)$ the same as $\forall y \ \exists x \ S(x,y)$?

Examples

Brothers are siblings

"Siblings" is symmetric

A mother is somebody's female parent

A first cousin is a child of a parent's sibling

Equality: Special predicate

Reflexive, transitive and symmetric

• Substituting equal objects doesn't change value of expressions

$$\forall x, y : x = y = 7 \left(S(x) \in S(y) \right)$$

- All models need to satisfy these properties
 - Typically, just assume that model has an "equality" relation,
 and the interpretation of the "=" symbol refers to that relation

Wumpus world in FOL

Modeling perception

$$\forall b, g, t \ Percept([Smell, b, g], t) \Rightarrow \underline{Smelt(t)}$$

$$\forall s, g, t \ Percept([s, Breeze, g], t) \Rightarrow Breezy(t)$$

Properties of locations

$$\forall x, t \; At(Agent, x, t) \land Smelt(t) \Rightarrow Smelly(x)$$

 $\forall x, t \; At(Agent, x, t) \land Breeze(t) \Rightarrow Breezy(x)$

Squares are breezy near a pit

$$\forall y \ Breezy(y) \Leftrightarrow \Big(\exists x \ Pit(x) \land Adjacent(x,y)\Big)$$

Modeling change

Facts hold only in certain situations

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Holding(Gold, \underline{t}) instead of Holding(Gold)
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- Can model using situation calculus
 - Add situation argument to each non-eternal predicate
 - ullet E.g., t in Holding(Gold,t)
- Model effects of actions using Result function
 Result(a,s) is the situation that results from doing a in s
- Effect axioms model changes due to actions

$$\forall s \ AtGold(s) \Rightarrow Holding(Gold, Result(Grab, s))$$

The frame problem

- In addition to modeling change, also need to model non-change
- Need "frame" axioms that model non-change

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\forall s \; HaveArrow(s) \Rightarrow HaveArrow(Result(Grab, s))
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- The frame problem:
 - Number of frame axioms can be large
 - Causes problems in inference

Solving the frame problem

- Successor state axioms:
 - For each *non-eternal* predicate, model how it is affected or not affected by actions
- P is true (=> [an action made P true or P is already true and no action made it false]
- Example for holding the gold

Planning using FOL

- Knowledge base (KB) contains all known facts
 - Successor state axioms
 - Properties of locations / perception
 - Initial conditions: $At(Agent, [1, 1], S_0)$ $Percept([NoSmell, Breeze, NoGlitter], S_0)$
- Use inference to find whether, e.g.,
 - Agent can obtain gold: $KB \vDash \exists s \; Holding(Gold, s)$ Example response:

$$\{s/Result(Grab, Result(Forward, S_0))\}$$

Agent can safely move to [2,2]

$$KB \vDash \exists s \ At(Agent, [2, 2], s) \land Alive(s)$$

Inference in FOL

Much more complicated then in propositional logic

- Two approaches
 - Propositionalization: Convert to propositional formula and use propositional inference
 - "Lifted" inference: Syntactically manipulate propositional sentences directly

Propositionalization

Create a propositional symbol for each atomic sentence

 Inferring propositional sentences by grounding universally quantified variables

Replace existential quantifier by introducing new constants

Problems with propositionalization

- May need to create lots of unnecessary symbols
- More importantly: Number of symbols could be infinite
 - If KB involves functions, can build infinitely many terms

 () If (k); Zero, Succ() = Odd (Zero), Odd (Succ (Zero)), Odd (Succ

Theorem (Herbrand '30)

If a sentence α is entailed by a FOL KB, then there is a proof using only a finite subset of the propositionalized KB

Naïve algorithm for FOL using propositionalization

- Want to determine whether sentence α is entailed by KB $\{i \in \mathcal{S}\}$
- Can enumerate all finite subsets PKB_1 , PKB_2 , ... of the propositionalized knowledge base $KB \land \neg \alpha$

For i = 1 to ∞

If PKB_i is unsatisfiable, e.g., using propositional resolution:

break and return *true*

- If $KB \models \alpha$ above algorithm stops after a finite number of steps
- If $KB \nvDash \alpha$ it will never stop
- This is intrinsic: FOL is semi-decidable

Lifted inference

Want to operate on FOL sentences directly

• Suppose we know $\forall x \ King(x) \land Greedy(x) \Rightarrow Evil(x)$ $King(John) \ \text{and} \ Greedy(John)$ $\theta = \{x/John\} \ \text{where} \ Evil(John)$

• Lifted inference allows to infer Evil(John) without instantiating

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

$$\uparrow \text{ king (Richard)} \land \text{ breedy (Richard)} \Rightarrow \text{ Evil(Richard)}$$

Generalized modus ponens

- Let p_1', \ldots, p_n' and p_1, \ldots, p_n be FOL sentences
- Suppose $\operatorname{Subst}(\theta,p_i') = \operatorname{Subst}(\theta,p_i)$ where θ is a substitution and $\operatorname{Subst}(\theta,p)$ its application

• E.g.:
$$\theta = \{ x / John \}$$
 $p'_i = King(x)$
 $p_i = king(Richard)$

Then Subst(θ, p_i') = $King(John)$ Subst(θ, p_i)= $king(Richard)$

Then the following inference is sound:

$$\frac{p'_1, \dots, p'_n, \quad p_1 \land \dots \land p_k}{\text{Subst}(\theta, q)} \Rightarrow q$$

Can use in a generalization of forward/backward chaining However, GMP is not complete

Generalization resolution

- Can also develop a lifted variant of resolution
- Details in reading
- Generalized resolution is sound and refutation-complete
 - ightharpoonup Can prove $KB \vdash_{GRes} \alpha$ if $KB \vDash \alpha$
 - ightharpoonup Cannot prove $KB \not\vdash_{GRes} \alpha$ if $KB \not\models \alpha$

Other logics

Language	Ontological commitment	Epistemological commitment
Propositional logic	facts	true/false/unknown
First order logic	facts, objects, relations	true/false/unknown
Higher order logic	facts, objects, relations, relations of relations,	true/false/unknown
Temporal logic	facts, objects, relations, times	true/false/unknown
Fuzzy logic	facts	degree of truth in [0,1]
Bayesian networks (up next)	facts	belief in probability of truth
Bayesian logic / Markov Logic Networks	facts, objects, relations	belief in probability of truth