Logic, Management & Frames

Dr. Collin F. Lynch

Al Academy: North Carolina State University

Copyright 2021 Collin F. Lynch



NC STATE

Agenda

First-Order Logic

Chaining

TMS

Knowledge Engineering



First-Order Logic



First Order Logic

- Set K of constant symbols: a, b, c, tweety, john, fido,
- Set V of variable symbols: T, U, V, W, X, Y,
- Set F of function symbols: F(), G(), H(), FatherOf(), FirstCat(),
- Set P of predicate symbols: p(), q(), r(), likes(), red(),
- ▶ Set of logical symbols: \land , \lor , \Rightarrow \Leftrightarrow , \neg
- Various punctuation: (), [], etc.
- Equality for variables Father(john) = henry
- ▶ Universal quantification $\forall X$
- Existential quantification ∃X

(Al Academy Dr. Lynch Optimization NC STATE UNIVERSITY 4/32

5/32

Example

```
1.has - feathers(tweety)
2.\forall X(has - feathers(X) \Rightarrow bird(X))
3.bird(tweety)
```

Dr. Lynch PAI Academy Optimization NC STATE UNIVERSITY

Models to Interpretations

- A set of objects U, called the universe of discourse
- ▶ A function $I_K : K \to U$ (i.e., for each constant symbol k in K, an element u in U)
- For each n-ary function symbol f in F, a function $I_f: U_n \to U$ (i.e., for each set of terms used as function arguments, a term interpreted as the return value)
- For each n-ary predicate symbol p in P, a function $I_p: P \rightarrow P*(Ux...(ntimes)...xU)$
- ► (P* is the powerset, the set of all subsets.)
- ▶ Alternatively, $I(p) \in U_n$ is defined by $I(p) = \{(t_1, ..., t_n) | p(t_1, ..., t_n) = True\}$, those sets of terms which make the predicate true.

Dr. Lynch Prairies Optimization NC STATE UNIVERSITY 6 / 32

Big Ideas

- 1. Convert to a logic problem (Propositionalization).
- 2. Handle Possible mappings (Unification).
- 3. Search (FC, BC & Resolution).

8 / 32

Example 1

Assume that Tweety has feathers. Assume that everything that has feathers is a bird. Prove that Tweety is a bird.

- Constants: tweety
- Variables: X
- ► Functions: none
- Predicates:
 - ▶ has feathers/1
 - ▶ bird/1

- 1.has feathers(tweety)
- $2.\forall X(has-feathers(X)\Rightarrow bird(X))$
- 3.bird(tweety)

Dr. Lynch Pal Academy Optimization NC STATE UNIVERSITY

Universal Instantiation

We can replace any universally instantiated variable in a statement with any ground term (g).

$$\frac{\forall v\alpha}{\mathsf{subst}(\{v/g\},\alpha)}\tag{1}$$

```
\forall x : cat(X) \Rightarrow stupid(X) \land hairy(X)
cat(tom) \Rightarrow stupid(tom) \land hairy(tom)
cat(X_{4214}) \Rightarrow stupid(X_{4214}) \land hairy(X_{4214})
cat(EnemyOf(Dog)) \Rightarrow stupid(EnemyOf(Dog)) \land hairy(EnemyOf(Dog))
```

(Al Academy Dr. Lynch Optimization NC STATE UNIVERSITY 9/32

Existential Instantiation (Skolem 1)

We replace existential variables with a new constant.

$$\frac{\exists v\alpha}{subst(\{v/C_{2121}\},\alpha)}$$

$$\exists x : cat(x) \land messy(x)$$

$$cat(C_{3311}) \land messy(C_{3311})$$

$$cat(C_{344354}) \land messy(C_{344354})$$

10/32

Dr. Lynch State UNIVERSITY

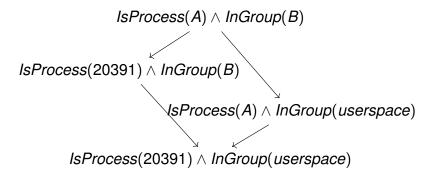
Generalized Modus Ponens (Lifted)

- 1.cat(sylvester)
- 2.bird(tweety)
- $3.cat(a) \land bird(b) \Rightarrow getsBeatenUpBy(a, b, t)$
- $4.\theta = \{a/sylvester, b/tweety\}$
- 5.getsBeatenUpBy(sylvester, tweety, t)

$$Unify(\alpha, \beta) = \theta) \Rightarrow Subst(\theta, \alpha) = Subst(\theta, \beta)$$

Dr. Lynch Prince Optimization NC STATE UNIVERSITY 11 / 32

Subsumption Lattice



(Al Academy Dr. Lynch Optimization **NC STATE UNIVERSITY** 12/32

Chaining



Forward Chaining

```
1.bird(opus)
```

- 2.bird(tweety)
- 3.yellow(tweety)
- 4.penguin(opus)
- 5.cat(sylvester)
- $6.cat(X) \Rightarrow hatesSnow(X)$
- $7.penguin(Y) \Rightarrow lovesSnow(Y)$
- $8.bird(Q) \Rightarrow flies(Q)$
- $9.(bird(Z) \land yellow(Z) \land cat(Q)) \Rightarrow chases(Q, Z)$

?chases(sylvester, tweety)

(Al Academy Dr. Lynch **NC STATE UNIVERSITY** Optimization 14 / 32

Forward Chaining

```
bird(opus)
2.
          bird(tweety)
3.
          yellow(tweety)
          penguin(opus)
5.
          cat(sylvester)
          cat(X) \Rightarrow hatesSnow(X)
          penguin(Y) \Rightarrow lovesSnow(Y)
8.
          bird(Q) \Rightarrow flies(Q)
          (flies(Z) \land yellow(Z) \land cat(Q)) \Rightarrow chases(Q, Z)
(1+8) \theta = \{Q/opus\} : flies(opus)
(2+8) \theta = \{Q/tweety\}: flies(tweety)
(4+7) \theta = \{Y/opus\} : lovesSnow(opus)
(5+6) \theta = \{X/sylvester\} : hatesSnow(sylvester)
```

Dr. Lynch RACKERY Optimization NC STATE UNIVERSITY 15 / 32

. . .

- 7. $penguin(Y) \Rightarrow lovesSnow(Y)$
- 8. $bird(Q) \Rightarrow flies(Q)$
- 9. $(flies(Z) \land yellow(Z) \land cat(Q)) \Rightarrow chases(Q, Z)$
- 10. $(1+8)\theta = \{Q/opus\}$: flies(opus)
- 11. $(2+8)\theta = \{Q/tweety\}$: flies(tweety)
- 12. $(4+7)\theta = \{Y/opus\} : lovesSnow(opus)$
- 13. $(5+6)\theta = \{X/sylvester\}$: hatesSnow(sylvester)
- 14. $(10+3+5+9)\theta = \{Z/tweety, Q/sylvester\}$ chases(sylvester, tweety)

Dr. Lynch PA Academy Optimization NC STATE UNIVERSITY 16 / 32

Heuristics

- Minimum Remaining Values: Fewest of X
- ▶ *Incremental FC*: Add one thing at a time.
- Goal Filtering: Filter irrelevant objects.
- Always NP-Hard.

Dr. Lynch PAIAcademy Optimization NC STATE UNIVERSITY 17 / 3

Backward Chaining

- bird(opus)
- 2. bird(tweety)
- 3. *yellow(tweety)*
- penguin(opus)
- 5. cat(sylvester)
- $cat(X) \Rightarrow hatesSnow(X)$
- 7. $penguin(Y) \Rightarrow lovesSnow(Y)$
- $bird(Q) \Rightarrow flies(Q)$
- 9. $(Flies(Z) \land yellow(Z) \land cat(Q)) \Rightarrow chases(Q, Z)$
- $(9)\theta = \{Z/sylvester, Q/opus\}$ chases(opus, sylvester)
- ? cat(opus)?

(Al Academy Dr. Lynch **NC STATE UNIVERSITY** Optimization 18/32

Backward Chaining

- 6. $cat(X) \Rightarrow hatesSnow(X)$
- 7. $penguin(Y) \Rightarrow lovesSnow(Y)$
- 8. $bird(Q) \Rightarrow flies(Q)$
- $(flies(Z) \land yellow(Z) \land cat(Q)) \Rightarrow chases(Q, Z)$
 - $(9)\theta = \overline{\{Z/opus, Q/sylvester\}}$ chases(sylvester, opus)?
- cat(sylvester)?
- 14. (5)*cat*(*sylvester*)
- yellow(opus)?

(Al Academy Dr. Lynch **NC STATE UNIVERSITY** Optimization 19/32

20 / 32

```
. .
```

- 6. $cat(X) \Rightarrow hatesSnow(X)$
- 7. $penguin(Y) \Rightarrow lovesSnow(Y)$
- 8. $bird(Q) \Rightarrow fies(Q)$
- 9. $(flies(Z) \land yellow(Z) \land cat(Q)) \Rightarrow chases(Q, Z)$
- ? $(9)\theta = \{Z/tweety, Q/sylvester\}$ chases(sylvester, tweety)?
- ? cat(sylvester)?
- 19. (5)*cat*(*sylvester*)
- ? yellow(tweety)?
- 21. (3) yellow(tweety)
- ? bird(tweety)?
- 23. (2) bird(tweety)...

TMS



Truth Maintenance Systems

- Allow KB updates with:
 - ▶ Tell(KB, α): explicit addition of knowledge.
 - ▶ Retract(KB, α): explicit retraction of knowledge.
- On each update we reevaluate the KB for potential changes.
- ▶ OR we store the justification with each entry:
 - We know Q because we know P and we know P implies Q.
 - $ightharpoonup \{P, P o Q\}$: Q
- We can also operationalize assumptions with entries.
- ► And we can generate **explanations**.

Dr. Lynch Prince Optimization NC STATE UNIVERSITY 22 / 32





23 / 32

Knowledge Engineering



What is Knowledge Engineering?



Problem Solving

- ► Real-World Problem Solving
 - Diagnose a patient.
 - Design a bridge.
 - Calculate your taxes.
- Problem Solving
 - Refine the problem to make it solvable.
 - Implement a solution.
 - Analyze the result.



Knowledge Engineering

- Knowledge Engineering
 - Building knowledge Based Systems
 - Designing knowledge bases.
 - Developing task analyses.
 - Curating information.
 - Structuring data.
- Closely related to:
 - Systems analysis.
 - Task analysis.
 - Cognitive Task Analysis

Dr. Lynch RACEdemy Optimization NC STATE UNIVERSITY 27 / 32

Knowledge Representation

- Capturing (human) knowledge in a form the computer can reason about.
- General CS Problem:
 - General solutions are data structures (e.g. Lists)
 - Al solutions are knowledge structures (e.g. Ontologies)
- Well-designed KR should:
 - Represent all the knowledge that is important to the problem.
 - Reflect the structure of the domain at the right granularity.
 - Support incremental development.
- Well-designed KR should not:
 - ▶ Be difficult (inefficient) to construct or use.
 - Require excess knowledge.

Dr. Lynch RARcotony Optimization NC STATE UNIVERSITY 28 / 32

Design problem: Modeling the <u>"right"</u> conditions, constraints, facts, etc. at the <u>"right"</u> level of detail in the <u>"usable"</u> way is hard!



Alternative Representations

- ► Logic & Predicate Calculus
- Rules & Production Systems
- Description Logics, Semantic Networks, Frames
- Scripts
- Ontologies



Alternative Representations

- Logic & Predicate Calculus
 - Very very rich representation (sound and complete).
 - Does not always match humans as we think.
 - Interface can be inefficient.
- Rules & Production Systems
 - Represent knowledge as if-then rules.
 - Inference can be quite efficient, natural, and modular.
 - Can even be non-monotonic.
 - Not necessarily sound or complete.
 - Conflict resolution required.
 - Inconsistencies create unintended consequences.

Dr. Lynch Prince Optimization NC STATE UNIVERSITY 31 / 32

Structured, Framed, & Associative Representations

- Represent the structure of the domain then reason on that
 - Belief networks
 - Frames
 - Scripts
 - Associative networks
- Basic structure:
 - Concepts (nodes)
 - Relationships between concepts (is-a, part, etc.).
 - Can also include: Inheritance. Procedural attachment.

(Al Academy Dr. Lynch Optimization **NC STATE UNIVERSITY** 32 / 32