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Horn Logics and Datalog

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Recap

- Looked at using Propositional Logic (PL) for representing knowledge
- effectively implementable (SAT solvers), but lacks expressiveness

Want a KR language that can ...

- 1. Represent sets of objects
- 2. Represent relationships between objects
- 3. Write statements that are true for some or all objects satisfying certain conditions
- 4. Express everything we can express in propositional logic (and, or, implies, not, ...)

→ First-order logic (FOL)

However: FOL satisfiability is undecidable

→ cannot hope for implementations





Propositional Horn Fragment

PL Horn Fragment: only allows the following kinds of formulas (n > 0):

$$P_1 \wedge \ldots \wedge P_n \rightarrow Q$$
 rules P facts

With P_i , Q being atoms, and where Q can be \perp .

Horn Clauses: Clauses with at most one positive literal.

$$\neg P_1 \lor \ldots \lor \neg P_n \lor Q$$

(Fact) Entailment. Input: set \mathcal{H} of Horn formulas and atom P Answer: true if every model of \mathcal{H} is also a model of P and false otherwise.

PL Horn entailment is solvable in polynomial time.





Lifting PL Horn to FOL Horn

First-Order Horn Clauses: Clauses with at most one positive literal But now, atoms can contain variables, constants, and function symbols. Implicit assumption: all variables ∀-quantified, but quantifiers not written Some examples of First-Order Horn clauses:

```
\neg JuvArthritis(x) \lor Arthritis(x)

\neg Arthritis(x) \lor \neg JuvDisease(x) \lor JuvArthritis(x)

\neg Child(x) \lor \neg Adult(x)

\neg Affects(x,y) \lor Person(y)

\neg JuvDisease(x) \lor Affects(x,f(x))

JuvDisease(JRA)
```





Horn Logics

Horn Formulas: FOL sentences whose CNF yield Horn clauses.

Horn Logics: Syntactic FOL fragments allowing only Horn Formulas.

Some examples of Horn formulas:

```
\forall x. (Arthritis(x) \land JuvDisease(x) \rightarrow JuvArthritis(x)) \\ \forall x. (Child(x) \land Adult(x) \rightarrow \bot) \\ \forall x. (\forall y. (Affects(x,y) \rightarrow Person(y))) \\ \forall x. (JuvDisease(x) \rightarrow \exists y. (Affects(x,y) \land Child(y))) \\ \forall x. (\forall y. (\forall z. (fatherOf(x,y) \land brotherOf(x,z) \rightarrow uncleOf(z,y)))) \\ JuvDisease(JRA)
```





Expressivity

We cannot express "disjunctive information":

Covering statements:

$$\forall x. (Person(x) \rightarrow Adult(x) \lor Child(x) \lor Teenager(x))$$

Negation on the left of implication

$$\forall x. (Person(x) \land \neg Woman(x) \rightarrow Man(x))$$

As well as many others ...

Note, however, that some formulas apparently "disjunctive" are Horn:

$$\forall x. (Adult(x) \lor Child(x) \lor Teenager(x) \rightarrow Person(x))$$





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Note, however, that some formulas apparently "disjunctive" are Horn:

$$\forall x. (Adult(x) \lor Child(x) \lor Teenager(x) \rightarrow Person(x))$$

... because they can be rewritten into formulas that are obviously Horn:

$$\forall x. (Adult(x) \rightarrow Person(x))$$

 $\forall x. (Child(x) \rightarrow Person(x))$
 $\forall x. (Teenager(x) \rightarrow Person(x))$





Existential Rules

$$\forall \vec{x}. \forall \vec{z}. (\varphi(\vec{x}, \vec{z}) \rightarrow \exists \vec{y}. \psi(\vec{x}, \vec{y})) \qquad \textit{Existential Rule} \\ \forall \vec{x}. \forall \vec{z}. (\varphi(\vec{x}, \vec{z}) \rightarrow \bot) \qquad \bot - \textit{Rule} \\ P(\vec{a}) \qquad \textit{Fact}$$

 $\varphi(\vec{x}, \vec{z})$: conjunction of function-free atoms with vars $\vec{x} \cup \vec{z}$. $\psi(\vec{x}, \vec{y})$: conjunction of function-free atoms with vars $\vec{x} \cup \vec{y}$.

$$\forall x. (Arthritis(x) \land JuvDisease(x) \rightarrow JuvArthritis(x))$$
 Rule $\forall x. (Child(x) \land Adult(x) \rightarrow \bot)$ \bot -Rule $\forall x. (JuvDisease(x) \rightarrow \exists y. (Affects(x,y) \land Child(y)))$ Rule $\exists uvDisease(JRA)$ Fact

Examples of Horn formulas outside this logic:

$$\forall x. (Adult(x) \lor Child(x) \lor Teenager(x) \rightarrow Person(x))$$





Reasoning with Existential Rules

Fact Entailment:

Input: pair $\langle \mathcal{R}, \mathcal{F} \rangle$ of rules and facts and a fact P. Answer is true iff $\langle \mathcal{R}, \mathcal{F} \cup \{\neg P\} \rangle$ is unsatisfiable.

Resolution (inferencing calculus for FOL) can be optimised for Horn clauses. General strategy: allow only certain kinds of resolution inferences:

- Need to show completeness
 Unsatisfiability must imply that the empty clause is derivable.
- No need to show soundness Still just resolution, which is sound.





Recall FOL Resolution Rule

$$\frac{\alpha \lor \varphi \quad \neg \beta \lor \psi}{(\varphi \lor \psi) MGU(\alpha, \beta)} \qquad \qquad \alpha, \beta \text{ are atoms}$$
 MGU(\alpha, \beta) is Most General Unifier of \alpha and \beta

Examples:

$$\frac{(\neg ArthritisPat(x) \lor Affects(f(x), x)) \quad ArthritisPat(g(a))}{Affects(f(g(a)), g(a))} \qquad \{x \mapsto g(a)\}$$

$$\frac{Affects(x, John) \quad \neg Affects(JRA, y)}{\Box} \qquad \{x \mapsto JRA, y \mapsto John\}$$

$$\frac{JuvDisease(h(g(f(x), a))) \quad \neg JuvDisease(h(g(y, y)))}{Rule\ not\ applicable}$$





Recall FOL Factoring Rule

$$\frac{y \vee \delta \vee \psi}{(y \vee \psi)MGU(y, \delta)} \quad y, \delta \text{ literals, same sign}$$

Examples:

$$\frac{ArthritisPat(x) \lor Affects(f(x), x) \lor ArthritisPat(g(a))}{Affects(f(g(a)), g(a)) \lor ArthritisPat(g(a))} \quad \{x \mapsto g(a)\}$$

$$\frac{Affects(x, John) \lor Affects(JRA, y)}{Affects(JRA, John)} \quad \{x \mapsto JRA, y \mapsto John\}$$

$$\frac{\neg JuvDisease(h(g(f(x), a))) \lor \neg JuvDisease(h(g(y, z)))}{\neg JuvDisease(h(g(f(x), a)))} \quad \{y \mapsto f(x), z \mapsto a\}$$





Recall FOL Resolution Procedure

```
    procedure Sat($)
    repeat
    for all clauses C<sub>1</sub>, C<sub>2</sub> in $ do
    $ := $ ∪ resolve(C<sub>1</sub>, C<sub>2</sub>)
    end for
    until No new clause can be added to $ or □ ∈ $
    If □ ∈ $ return false
    return true
    end procedure
```

Function resolve(C_1 , C_2) applies FO resolution in all possible ways, and then applies factoring in all possible ways.





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```

Function resolve(C_1 , C_2) applies FO resolution in all possible ways, and then applies factoring in all possible ways.

Wait ... in all possible ways?





Resolution with free selection: a complete strategy

- Calculus parameterised by a Selection Function S
- S assigns to each Horn clause C a non-empty subset of its literals:
 - S(C) contains the single positive literal, OR
 - *S*(*C*) contains a subset of negative literals
- Restrict resolution such that we only resolve on selected literals

We are free to design the selection function ourselves:

If we satisfy the basic constraints, completeness is guaranteed.





A reasonable selection function:

- Select the set of all negative literals in each clause
- If there is no negative literal, select the (unique) positive literal

As usual, we prepare for resolution (Skolemisation, CNF, clause form)

$$A(x) \to \exists y. R(x, y) \land B(y) \quad \rightsquigarrow \quad \neg A(x) \lor R(x, f(x))$$

$$\neg A(x) \lor B(f(x))$$

$$B(x) \to C(x) \quad \rightsquigarrow \quad \neg B(x) \lor C(x)$$

$$R(x, y) \land C(y) \to D(x) \quad \rightsquigarrow \quad \neg R(x, y) \lor \neg C(y) \lor D(x)$$

$$A(a) \quad \rightsquigarrow \quad A(a)$$

We now want to see whether D(a) follows ...





$$\neg A(x) \lor R(x, f(x)) \tag{1}$$

$$\neg A(x) \lor B(f(x)) \tag{2}$$

$$\neg B(x) \lor C(x) \tag{3}$$

$$\neg R(x, y) \lor \neg C(y) \lor D(x) \tag{4}$$

$$A(a) \tag{5}$$

$$\neg D(a) \tag{6}$$

With this selection, we don't need to resolve (1) and (4) Observation: This strategy amounts to Unit Resolution One of the premises of resolution must be a unit clause!





 $A(x) \setminus D(x, f(x))$

$\neg A(X) \lor R(X,f(X))$		(1)
$\neg A(x) \lor B(f(x))$		(2)
$\neg B(x) \lor C(x)$		(3)
$\neg R(x,y) \lor \neg C(y) \lor D(x)$		(4)
A(a)		(5)
$\neg D(a)$		(6)
R(a, f(a))	(1) + (5)	(7)
B(f(a))	(2) + (5)	(8)
C(f(a))	(8) + (3)	(9)
$\neg C(f(a)) \lor D(a)$	(7) + (4)	(10)
D(a)	(9) + (10)	(11)
	(11) + (6)	(12)





111

We still have termination problems ...

$$A(x) \rightarrow \exists y. R(x, y) \land A(y) \quad \leadsto \quad \neg A(x) \lor R(x, f(x))$$

$$\neg A(x) \lor A(f(x))$$

$$A(a) \quad \leadsto \quad A(a)$$

 $\neg A(x) \lor R(x, f(x))$ $\neg A(x) \lor A(f(x))$ A(a) R(a, f(a)) A(f(a)) R(a, f(f(a)) A(f(f(a)))



We still have termination problems		$\neg A(x) \lor R(x, f(x))$
		$\neg A(x) \lor A(f(x))$
$A(x) \rightarrow \exists y. R(x,y) \land A(y) \rightsquigarrow$	$\neg A(x) \lor R(x, f(x))$	A(a)
	$\neg A(x) \lor A(f(x))$	R(a, f(a))
$A(a) \rightsquigarrow$	A(a)	A(f(a))
		R(a, f(f(a)))
		A(f(f(a))

Theorem

Unsatisfiability and fact entailment over existential rules are undecidable (semi-decidable).

That is, as difficult as checking unsatisfiability in FOL.





Datalog

To achieve decidability we need to sacrifice (even more) expressivity.

Datalog: The quintessential rule-based KR language

$$\forall \vec{x}. \forall \vec{z}. (\varphi(\vec{x}, \vec{z}) \to \psi(\vec{x})) \qquad Rule$$

$$\forall \vec{x}. \forall \vec{z}. (\varphi(\vec{x}, \vec{z}) \to \bot) \qquad \bot -Rule$$

$$P(\vec{a}) \qquad Fact$$

 $\varphi(\vec{x}, \vec{z})$ and $\psi(\vec{x})$: conjunctions of function-free atoms

We can still express

$$\forall x.(\forall y.(\forall z.(fatherOf(x,y) \land brotherOf(x,z) \rightarrow uncleOf(z,y))))$$

 $\forall x.(\forall y.(Affects(x,y) \rightarrow Person(y)))$

But, we can no longer express

$$\forall x.(JuvDisease(x) \rightarrow \exists y.(Affects(x,y) \land Child(y))))$$





Decidability of Entailment

Theorem

Fact entailment in Datalog is decidable.

Decidability follows directly from Herbrand's theorem

- Our problem reduces to unsatisfiability of $S = \Re \cup \Im \cup \{\neg P\}$
- $\mathcal{R} \cup \mathcal{F} \cup \{\neg P\}$ is a set of clauses without function symbols so Herbrand universe finite
- Gilmore's FOL unsatisfiability algorithm terminates.





Decidability of Entailment

Our algorithm is an adaptation of Gilmore's when Herbrand universe is finite

```
1: procedure Datalog-Gil(\langle \mathcal{R}, \mathcal{F} \rangle, P)
```

- 2: Compute Herbrand Universe *U*
- 3: $\mathbb{R}' := \operatorname{ground}(\mathbb{R}, U)$
- 4: **return** Horn-Prop($\langle \mathcal{R}', \mathcal{F} \rangle$, *P*)
- 5: end procedure

Subroutine Horn-Prop solves entailment problem for Horn PL





Complexity Considerations

```
\forall x. (\forall y. (\forall z. (fatherOf(x, y) \land brotherOf(x, z) \rightarrow uncleOf(z, y))))
fatherOf(John, Mary)
brotherOf(John, Peter)
```

Herbrand universe: constants in $\langle \mathcal{R}, \mathcal{F} \rangle$

$$U = \{John, Mary, Peter\}$$

Grounding leads to exponential size set of propositional clauses

```
fatherOf(John, John) \land brotherOf(John, John) \rightarrow uncleOf(John, John)

fatherOf(John, Mary) \land brotherOf(John, Mary) \rightarrow uncleOf(Mary, Mary)

fatherOf(John, Peter) \land brotherOf(John, Peter) \rightarrow uncleOf(Peter, Peter)

and so on
```

Size of the grounding grows as $O(c^{\nu})$, where

- c is the max. number of constants in facts.
- v is the max. number of variables in rules.





Complexity Considerations

Propositional entailment in Horn PL can be decided in polynomial time. Overall process takes exponential time (because of grounding).

Theorem

Fact entailment in Datalog is decidable in ExpTime.

In fact, the problem is also ExpTime-hard (beyond this course).

→ Naive grounding algorithm is worst-case optimal.





Practical Considerations

From a practical point of view, we can do much better:

- · Avoid computing the grounding upfront
- Instantiate variables to constants "on the fly"

We develop two resolution-based strategies:

1. Forward chaining:

Start from facts and instantiate rules to derive new facts whenever possible until goal is derived

2. Backward chaining:

Start from goal and proceed "backwards" to derive the empty clause

Both strategies can be seen as Resolution with Free Selection.





Forward Chaining (Example)

Start from facts and instantiate rules to derive new facts whenever possible until goal (or \square) is derived.

Example: Check if *Child(John*) follow from the following:

$$\forall x.(JuvArthritis(x) \rightarrow JuvDisease(x))$$
 (13)

$$\forall x. (\forall y. (JuvDisease(x) \land Affects(x, y) \rightarrow Child(y))) \tag{14}$$

$$JuvArthritis(JRA)$$
 (15)

$$Affects(JRA, John)$$
 (16)

Match existing facts to rule bodies to derive new facts.

From Fact (15) and Rule (13) we obtain the following by unit resolution

$$JuvDisease(JRA)$$
 (17)

From Facts (17) and (16) and Rule (14), derive goal and stop.

Child(John)





Forward Chaining and Resolution

 S_{fw} : select all negative literals in clauses, and the (unique) positive literal if the clause does not have negative literals.

¬JuvArthritis(x) ∨ JuvDisease(x) JuvArthritis(JRA)

We obtain the following by resolution:

JuvDisease(JRA)





Forward Chaining and Resolution

 S_{fw} : select all negative literals in clauses, and the (unique) positive literal if the clause does not have negative literals.

Deriving a new fact by matching other facts to a rule may require several resolution steps (Hyperresolution).

```
\neg JuvDisease(x) \lor \neg Affects(x, y) \lor Child(y)
Affects(JRA, John)
JuvDisease(JRA)
```

We obtain the following by resolution:

```
¬JuvDisease(JRA) \lor Child(John)
Child(John)
```

In forward chaining, we do both steps in one.





Forward Chaining

```
1: procedure Forward(\langle \mathcal{R}, \mathcal{F} \rangle, P)
           \mathfrak{F}' := \mathfrak{F}
 3:
           repeat
 4.
                 for each rule R = \neg B_1 \lor \neg B_2 \lor \dots, \lor \neg B_n \lor H \in \mathbb{R} do
                      if \{D_1, \ldots, D_n\} \subseteq \mathcal{F}' such that B_i unifies with D_i then
 5:
                            \theta := \text{Unify}(\{B_1 \doteq D_1, \dots, B_n \doteq D_n\})
 6:
                            \mathfrak{F}' := \mathfrak{F}' \cup \{H\theta\}
                      end if
 8.
 9:
                 end for
           until No new atom can be added to \mathfrak{F}' or P \in \mathfrak{F}' or \square \in \mathfrak{F}'
10.
           if P \in \mathcal{F}' or \square \in \mathcal{F}' then
11:
12:
                 return true
13.
           else
14.
                 return false
           end if
15.
16: end procedure
```





Backward Chaining (Example)

Check whether following rules and facts imply *Child*(*John*):

$$\forall x.(JuvArthritis(x) \rightarrow JuvDisease(x)) \tag{18}$$

$$\forall x. (\forall y. (JuvDisease(x) \land Affects(x, y) \rightarrow Child(y))) \tag{19}$$

$$JuvArthritis(JRA)$$
 (20)

$$Affects(JRA, John)$$
 (21)

Match "goal" *Child*(*John*) to rule heads and facts to derive new goals.

To prove *Child(John*), by Rule (19) it is sufficient to show

Then, by Fact (21), it would be sufficient to show *JuvDisease*(*JRA*). Another possibility is to use Rule (18) and get the following sub-goals

JuvArthritis(x) and *Affects*(x, *John*)

And so on ...





Backward Chaining (Example)

We can represent this kind of backwards reasoning in an AND-OR tree:

```
\forall x.(JuvArthritis(x) \rightarrow JuvDisease(x))
\forall x.(\forall y.(JuvDisease(x) \land Affects(x,y) \rightarrow Child(y)))
JuvArthritis(JRA)
Affects(JRA,John)
```

```
Child(John)

JuvDisease(x), Affects(x, John)

JuvDisease(JRA) JuvArthritis(x), Affects(x, John)

JuvArthritis(JRA) Affects(JRA, John) ...
```





Backward Chaining and Resolution

 S_{bw} : select the unique positive literal in clauses, and all negative literals if the clause does not have positive literals.

Matching the goal to a rule head or a fact corresponds to one resolution step.

$$\frac{\neg JuvDisease(x) \lor \neg Affects(x,y) \lor Child(y) \quad \neg Child(John)}{\neg JuvDisease(x) \lor \neg Affects(x,John)}$$





Termination Issues

Resolution with free selection may not terminate with S_{bw} .

Example: Show that John is a Scientist.

$$\neg worksWith(x,y) \lor \neg Scientist(y) \lor Scientist(x)$$
 (22)
 $worksWith(John, Mary)$ (23)
 $\neg Scientist(John)$ (24)

We start resolving on selected atoms:

$$\neg worksWith(John, y) \lor \neg Scientist(y)$$
 (22) + (24) (25)

$$\neg worksWith(John, y_1) \lor \neg worksWith(y_1, y_2) \lor \neg Scientist(y_2)$$
 (22) + (25) (26)

. . .

Keep on generating clauses with chains of *worksWith* atoms of increasing length (variable proliferation).

Thus, the backward chaining tree can have infinite branches.





Other Considerations

Implementing Forward and Backward chaining efficiently is non-trivial:

- Forward chaining: set of deduced facts might get huge
- Backward chaining: recursion may be too deep or search tree too wide.

There are many ways to optimise these algorithms Semi-naive evaluation, Magic sets, ...

But, this is beyond the scope of this course.

There are many optimised systems that implement forward/backward chaining.

The KR languages we have described are related to:

- Databases: Datalog query language, and deductive databases
- Logic programming: Prolog



