## What is Database Theory?

A collection of studies, often connected to the relational model of data.

- Restricted forms of logic, between SQL and full first-order.
- Dependency theory: generalizing functional dependencies.
- Conjunctive queries (CQ's): useful, decidable special case.
- "Universal relations" fitting a database schema into a single (virtual) relation.

#### Why Care?

A lot of this work was, quite frankly, done "for the fun of it." However, it turns out to have unexpected applications, as natural ideas often do:

- Information integration:
  - Logic, CQ's, etc., used for expressing how information sources fit together.
  - Recent work using universal-relation too — eliminates requirement that user understand a lot about the integrated schema.
- More powerful query languages.
  - Recursion needed in repositories, other applications.
  - Database logic provided some important ideas used in SQL3 standard: seminaive evaluation, stratified negation.
- Potential application: constraints and triggers are inherently recursive. When do they converge?

## **Outline of Topics**

- Logic intro, especially logical rules (if-then), 1. dealing with negation.
  - In database logic there is a special semantics frowned upon by Mathematicians, but it works.
- Logic processing: optimizing collections of 2.rules that constitute a query.

- "Magic-sets" technique for recursive queries.
- 3. Conjunctive queries: decidability of containment, special cases.
- Information-integration architectures: rule 4. expansion vs. systems that piece together solutions to queries from logical definitions of sources.
  - Important CQ application.
- Universal relation data model: answering 5.queries without knowing the schema.

<sup>♣</sup> 

- 6. Other stuff if I have time for it and/or there is class interest:
  - a) Data mining of databases.
  - b) Materialized views, warehouses, data cubes.

# **Course Requirements**

- 1. The usual stuff: midterm, final, problem sets.
- 2. A project:
  - Each student should attempt to implement an algorithm for one of the problems discussed in the class.
  - Your choice, but you should pick something that is combinatorially hard, i.e., the problem is dealing efficiently with large cases.
  - ✤ I'll suggest some problems as we go, and keep a list on the Web page.

## **Review of Logic as a Query Language**

Datalog programs are collections of rules, which are Horn clauses or if-then expressions.

## Example

The following rules express what is needed to "make" a file. It assumes these relations or EDB (*extensional database*) predicates are available:

- 1. source(F): F is a source file, i.e., stored in the file system.
- 2. includes(F,G): file F includes file G.
- 3. create(F, P, G): we create file F by applying process P to file G.

req(F,F) :- source(F)
req(F,G) :- includes(F,G)
req(F,G) :- create(F,P,G)
req(F,G) :- req(F,H) & req(H,G)

#### Rules

#### Head :- Body

- :- is read "if"
- Atom = predicate applied to arguments.
- Head is atom.
- Body is logical AND of zero or more atoms.
- Atoms of body are called *subgoals*.
- Head predicate is IDB *intensional database* = predicate defined by rules. Body subgoals may have IDB or EDB predicates.
- Datalog program = collection of rules. One IDB predicate is distinguished and represents result of program.

#### Meaning of Rules

The head is true for its arguments whenever there exist values for any *local* variables (those that appear in the body, but not the head) that make all the subgoals true.

#### Extensions

1. Negated subgoals. Example:

cycle(F) :- req(F,F) & NOT source(F)

2. Constants as arguments. Example:

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req(F,"stdio.h") :- type(F,"cCode")
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- 3. Arithmetic subgoals. Example: composite(A) :- divides(B,A) & B > 1 & B <> A

Applying Rules ("Naive Evaluation") Given an EDB:

- 1. Start with all IDB relations empty.
- 2. Instantiate (with constants) variables of all rules in all possible ways. If all subgoals become true, then infer that the head is true.
- 3. Repeat (2) in "rounds," as long as new IDB facts can be inferred.
- (2) makes sense and is finite, as long as rules are *safe* = each variable that appears anywhere in the rule appears in some nonnegated, nonarithmetic subgoal of the body.
- Limit of (1)-(3) = Least fixed point of the rules and EDB.

#### Seminaive Evaluation

- More efficient approach to evaluating rules.
- Based on principle that if at round i a fact is inferred for the first time, then we must have used a rule in which one or more subgoals were instantiated to facts that were inferred on round i - 1.

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Thus, for each IDB predicate p, keep both relation P and relation  $\Delta P$ ; the latter represents the new facts for p inferred on the most recent round.

## Outline of SNE Algorithm

- 1. Initialize IDB relations by using only those rules without IDB subgoals.
- 2. Initialize the  $\Delta$ -IDB relations to be equal to the corresponding IDB relations.
- 3. In one round, for each IDB predicate p:
  - a) Compute new  $\Delta P$  by applying each rule for p, but with *one* subgoal treated as a  $\Delta$ -IDB relation and the others treated as the correct IDB or EDB relation. (Do for *all* possible choices of the  $\Delta$ -subgoal.)
  - b) Remove from new  $\Delta P$  all facts that are already in P.
  - c)  $P := P \cup \Delta P$ .
- 4. Repeat (3) until no changes to any IDB relation.

#### Example

- (1) req(F,F) :- source(F)
  (2) req(F,G) :- includes(F,G)
  (3) req(F,G) :- create(F,P,G)
  (4) req(F,G) :- req(F,H) & req(H,G)
- Assume EDB relations S, I, C and IDB relation R, with obvious correspondence to predicates.
- Initialize:  $R := \Delta R := \sigma_{\#1=\#2}(S \times S) \cup I \cup \pi_{1,3}(C).$

• Iterate until 
$$\Delta R = \emptyset$$
:

- 1.  $\Delta R := \pi_{1,3}(R \bowtie \Delta R \cup \Delta R \bowtie R)$
- $2. \quad \Delta R := \Delta R R$
- 3.  $R := R \cup \Delta R$

# Models

Model of rules + EDB facts = set of atoms selected to be true such that

- 1. An EDB fact is selected true iff it is in the given EDB relation.
- 2. All rules become true under any instantiation of the variables.
  - Facts not stated true in the model are assumed false.
  - Only way to falsify a rule is to make each subgoal true and the head false.
- *Minimal model* = model + no proper subset is a model.
- For a Datalog program with only nonnegated, relational atoms in the bodies, the *unique* minimal model is what naive or seminaive evaluation produces, i.e., the IDB facts we are *forced* to deduce.
- Moreover, this LFP is reached after a finite number of rounds, if the EDB is finite.

# **Function Symbols**

Terms built from

- 1. Constants.
- 2. Variables.
- 3. Function symbols applied to terms as arguments.
  - ◆ Example:

addr(street(maple), number(101))

#### Example

Binary trees defined by

isTree(null)
isTree(node(L,T1,T2)) : label(L) &
 isTree(T1) &
 isTree(T2)

If label(a) and label(b) are true, infers facts like isTree(node(a, null, null))isTree(node(b, null, node(a, null, null)))

- Application of rules as for Datalog: make all possible instantiations of variables and infer head if all subgoals are true.
- LFP is still unique minimal model, as long as subgoals are relational, nonnegated.
- But LFP may be reached only after an infinite number of rounds.