# **Rewriting preferences as queries**

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## fact:

when you query a big data set chances are that the answer will be big as well

# problem:

it is difficult for the user to find an item of interest in a big answer set

# one solution:

partition the answer into smaller sets and show one part at a time

(showing "best parts" first)

 $\rightarrow$  one way to achieve this is to exploit user preferences

# various kinds of user preferences

# • in terms of their nature preferences can be:

quantitative (John likes BMWs 80%, and VWs 60%, ...)

**qualitative** (I like BMW more than VW) → easy to express by the casual user

# • in terms of their persistence in time preferences can be:

#### long term preferences

either discovered by the system (unobtrusively, from query logs) or declared explicitly by the user

#### short term preferences

expressed explicitly by the user, online

• the nature and persistence in time are orthogonal features of preferences

• in this presentation we focus on qualitative preferences (whether short or long term)

# the basic setting through an example

Serial	Make	Color	Mileage	Price	Year
1	BMW	black	35000	3800	2002
2	Honda	blue	63000	2900	2000

 Table T (e-catalogue such as Autoreflex)

as T is usually large chances are that query answers are large as well

**usual query Q: VW**v**BMW** (meaning: select \* from T where (Make=VW)v(Make=BMW)

preference P: Red → Black (meaning "Red is preferred to Black")

#### answer:

**standard approach:** compute ans(Q) then use P to partition ans(Q) into a sequence of two tables:  $T_{Red}$ ,  $T_{Black}$ **rewriting approach**: use P to rewrite Q into a sequence of two queries:  $Q_{Red} \land Q$ ,  $Q_{Black} \land Q$  whose answers are  $T_{Red}$ ,  $T_{Black}$ 



#### ordered partition of ans(Q)

#### advantage of rewriting :

rewriting is transparent to the system, it is possible to do incremental evaluation under user control

the problem we consider: rewrite P into the sequence Q<sub>0</sub>, ..., Q<sub>m</sub> (forgetting about Q for the moment, i.e. P <u>is</u> the query)

ICDE 2008, ADBIS 2011, ASSETS Project (2010-2012), ICFCA 2012,

#### preference over attribute A : pair of values (v, w) from the domain of A ( $v \neq w$ ) (read as "v is preferred to w" or "v precedes w", or "v dominates w", denoted as $v \rightarrow w$ )

preference relation or preference graph over A : a set P.A of preferences over A



assumption : the only values of interest to the user are those expressed in the preference graph (hence a sort of "closed world assumption") as a consequence, the only tuples of interest are those in the answer to the following query: *induced query:* Q(P.A)= disjunction of all values in P.A in our example: Q(P.Make)= BMW\Toyota\VW\Honda

#### no other assumption is made on the preference relation

(in particular, the preference relation may not be transitive and may contain cycles)

# ranking of nodes in the preference graph :

(assuming acyclicity for now and we'll see shortly how we can handle cycles)

*if* v is a root *then* rank(v)=0

else rank(v)= the maximal length of path among all paths going from a root to v



- the further away from the roots the less preferred a value is
- each non root value of rank i>0 is preceded by a value of rank i-1

# rewriting a preference graph into a sequence of queries based on ranks:



• each query Q<sub>i</sub> is evaluated in order of increasing rank (hence best answers first)

- evaluation is under user control (incremental evaluation)
- result presentation adapts to user preferences

(*i.e.* the answer to the same induced query will be presented differently for different preferences) • skyline of T:

it is defined to be the set of all non dominated tuples of T (w.r.t. the given preferences), hence it is the answer to the first query  $Q_i$  such that ans( $Q_i$ )#Ø

(and this Q<sub>i</sub> depends on the current instance of T)

# additional issues

#### user profile:

the sequence Q<sub>0</sub>, Q<sub>1</sub>, Q<sub>2</sub>, ..., Q<sub>m</sub> into which the preference graph is rewritten can be seen as the user's prefrence profile (to be stored, and invoked in subsequent query sessions)

# **Q**<sub>i</sub> 's with large answer sets:

**Q**<sub>i</sub> can be modified using order-by and/or top k and/or "hard conditions" (eventually through a dialogue with the user)

### Ex:

Q : select \* from T where (P.Make: BMW→VW) order-by Km is rewritten as: Q0: select \* from T where (Make=BMW) order-by Km Q1: select \* from T where (Make=VW) order-by Km

# handling cycles in the preference graph

with the help of the user : show the cycles and ask the user to 'break' them
 → need for algorithms for finding all cycles in the preference graph

### 2. without user involvement:

- 2.1 refusing the addition of a preference if the current acyclic graph becomes cyclic
- 2.2 considering all values in a cycle to be of equal preference, thus turning each maximal cycle into one node
  - $\rightarrow$  need for algorithms for finding all cycles in the preference graph



# extending the approach

preferences over more than one attribute can be declared either <u>conjunctively</u> or <u>independently</u>

in a conjunctive declaration we proceed as in the case of a single attribute (the only difference is that the nodes are conjunctions of values instead of single values)

ex: **P.{Make, Colour}** = {VW^Red > VW^Black, BMW^Red > BMW^Black}



### **Step 1:** combine the values present in the preference relations:



**Step 2:** define a preference rel. over the combined values based on the given preference rel.

to perform this second step we need a preference rel. over the attributes (given by the user)

Pareto: all attributes are of the same importance Prioritized: there is a prioritiy over attributes

(priority: linear ordering of the attributes over which preferences are declared)

**Preferences : P.Make**= {VW→BMW}, **P.Colour**= {Red→Black}



once the graph is constructed we proceed to rewriting as for single attributes: Pareto:  $B_0 = \{VW \land Red\}, B_1 = \{VW \land Black, BMW \land Red\}, B_2 = \{BMW \land Black\}$ Prioritized:  $B_0 = \{VW \land Red\}, B_1 = \{VW \land Black\}, B_2 = \{BMW \land Red\}, B_3 = \{BMW \land Black\}$ 

*question: can we find the*  $B_0$ ,  $B_1$ , .... *without constructing the derived graphs?* (*i.e. based on the graphs of the individual attributes*)

# avoiding derived graphs:

P.Make= { $VW \rightarrow BMW$ } $B_0^M = {VW}$ ,  $B_1^M = {BMW}$ P.Colour= { $Red \rightarrow Black$ } $B_0^C = {Red}$ ,  $B_1^C = {Black}$ 

Prioritized : (Make→ Colour)

 $B_0 = B_0^M x B_0^C = \{VW \land Red\}$   $B_1 = (B_0^M x B_1^C) = \{VW \land Black\}$   $B_2 = B_1^M x B_0^C) = \{BMW \land Red\}$  $B_3 = B_1^M x B_1^C) = \{BMW \land Black\}$ 

**skyline**: the answer of the first query  $Q_i$  such that  $ans(Q_i) \neq \emptyset$ 

Pareto :  $B_0 = B_0^{M} \times B_0^{C} = \{VW \land Red\}$  $B_1 = (B_0^{M} \times B_1^{C}) \cup (B_1^{M} \times B_0^{C}) = \{VW \land Black, BMW \land Red\}$  $B_2 = B_1^{M} \times B_1^{C}) = \{BMW \land Black\}$ 

skyline: the answer of the first query Q<sub>i</sub> such that ans(Q<sub>i</sub>)≠Ø is a <u>subset</u> of the skyline , not necessarily the whole skyline (because of the unions)

**1:** 
$$a \rightarrow b \rightarrow c$$
  
**2:**  $x \rightarrow y \rightarrow z$   
**3:**  $a \rightarrow b \rightarrow c$   
**4:**  $B_0^1 = \{a\}, B_1^1 = \{b\}, B_2^1 = \{c\}$   
**5:**  $B_0^2 = \{x\}, B_1^2 = \{y\}, B_2^2 = \{z\}$ 

## Pareto over 1, 2:

 $Q_{0} = a \wedge x$   $Q_{1} = (a \wedge y) \vee (b \wedge x)$   $Q_{2} = (a \wedge z) \vee (b \wedge y) \vee (c \wedge x)$   $Q_{3} = (b \wedge z) \vee (c \wedge y)$   $Q_{4} = c \wedge z$   $a \wedge y$   $a \wedge y$   $a \wedge y$   $b \wedge y$   $b \wedge y$   $b \wedge z$   $c \wedge y$   $b \wedge z$   $c \wedge y$ 

assuming ans(a $\land$ x)= $\varnothing$  and ans(b $\land$ x)= $\varnothing$ ,

the skyline is  $ans(a \land y) \cup ans(c \land x)$ 

- skyline can be computed incrementally (under user control)
- can be used to compute skylines over joins without computing the join

# summarizing

preference query= usual query + preferences + priority + skyline + .....



# concluding remarks

- attribute values organized in a taxonomy (domains of relational tables are unordered sets)
- do we need to impose conditions on preference relations (e.g. transitivity)
- do we need more general kinds of priorities?
   (e.g. priority over attribute sets, each set assumed to be Pareto)
- what happens with attributes over which no preferences are expressed don't care assumption? (as usual in databases), ceteris paribus assumption? a preference specification language for expressing preferences and conditions?
- **combining usual queries and preferences** (priority to the query? to the preferences? no priority?)

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