

Lecture 23: Query Optimization (3)

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Outline

- Search space
- Algorithms for enumerating query plans
- Estimating the cost of a query plan

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Computing the Cost of a Plan

- Collect statistical summaries of stored data
- Estimate size in a bottom-up fashion
- Estimate cost by using the estimated size

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Statistics on Base Data

- Collected information for each relation
 - Number of tuples (cardinality)
 - Indexes, number of keys in the index
 - Number of physical pages, clustering info
 - Statistical information on attributes
 - Min value, max value, number distinct values
 - Histograms
 - Correlations between columns (hard)
- Collection approach: periodic, using sampling

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Size Estimation Problem

```
S = SELECT list  
      FROM   R1, ..., Rn  
      WHERE cond1 AND cond2 AND ... AND condk
```

Given T(R1), T(R2), ..., T(Rn)
Estimate T(S)

How can we do this ? Note: doesn't have to be exact.

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Size Estimation Problem

```
S = SELECT list  
      FROM   R1, ..., Rn  
      WHERE cond1 AND cond2 AND ... AND condk
```

Remark: $T(S) \leq T(R1) \times T(R2) \times \dots \times T(Rn)$

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Selectivity Factor

- Each condition *cond* reduces the size by some factor called selectivity factor
- Assuming independence, multiply the selectivity factors

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Example

R(A,B)
S(B,C)
T(C,D)

```
SELECT *  
FROM R, S, T  
WHERE R.B=S.B and S.C=T.C and R.A<40
```

T(R) = 30k, T(S) = 200k, T(T) = 10k

Selectivity of R.B = S.B is 1/3

Selectivity of S.C = T.C is 1/10

Selectivity of R.A < 40 is ½

What is the estimated size of the query output ?

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Rule of Thumb

- If selectivities are unknown, then:
selectivity factor = 1/10
[System R, 1979]

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Selectivities from Statistics

- Condition is $A = c$ /* value selection on R */
– Selectivity = $1/V(R,A)$
- Condition is $A < c$ /* range selection on R */
– Selectivity = $(c - \text{Low}(R, A)) / (\text{High}(R, A) - \text{Low}(R, A))T(R)$
- Condition is $A = B$ /* $R \bowtie_{A=B} S$ */
– Selectivity = $1 / \max(V(R,A), V(S,A))$
– (will explain next)

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Assumptions

- Containment of values: if $V(R,A) \leq V(S,B)$, then the set of A values of R is included in the set of B values of S
 - Note: this indeed holds when A is a foreign key in R, and B is a key in S
- Preservation of values: for any other attribute C, $V(R \bowtie_{A=B} S, C) = V(R, C) \text{ (or } V(S, C)\text{)}$

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Selectivity of $R \bowtie_{A=B} S$

Assume $V(R,A) \leq V(S,B)$

- mmmmmmmhhk,mmmbknmmmmmmmmmmmmkktt
- Each tuple t in R joins with $T(S)/V(S,B)$ tuple(s) in S
- Hence $T(R \bowtie_{A=B} S) = T(R) T(S) / V(S,B)$

In general: $T(R \bowtie_{A=B} S) = T(R) T(S) / \max(V(R,A), V(S,B))$

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Size Estimation for Join

Example:

- $T(R) = 10000, T(S) = 20000$
- $V(R,A) = 100, V(S,B) = 200$
- How large is $R \bowtie_{A=B} S$?

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Histograms

- Statistics on data maintained by the RDBMS
- Makes size estimation much more accurate (hence, cost estimations are more accurate)

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Histograms

Employee(ssn, name, age)

$T(\text{Employee}) = 25000, V(\text{Employee}, \text{age}) = 50$
 $\min(\text{age}) = 19, \max(\text{age}) = 68$

$\sigma_{\text{age}=48}(\text{Employee}) = ?$ $\sigma_{\text{age}>28 \text{ and } \text{age}<35}(\text{Employee}) = ?$

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Histograms

Employee(ssn, name, age)

$T(\text{Employee}) = 25000, V(\text{Employee}, \text{age}) = 50$
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Estimate = $25000 / 50 = 500$ Estimate = $25000 * 6 / 60 = 2500$

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Histograms

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Age:	0..20	20..29	30-39	40-49	50-59	> 60
Tuples	200	800	5000	12000	6500	500

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Histograms

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Age:	0..20	20..29	30-39	40-49	50-59	> 60
Tuples	200	800	5000	12000	6500	500

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Estimate = 1200 Estimate = $2*80 + 5*500 = 2660$

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Types of Histograms

- How should we determine the bucket boundaries in a histogram ?

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Types of Histograms

- How should we determine the bucket boundaries in a histogram ?
- Eq-Width
- Eq-Depth
- Compressed

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Employee(ssn, name, age)

Histograms

Eq-width:

Age:	0..20	20..29	30-39	40-49	50-59	> 60
Tuples	200	800	5000	12000	6500	500

Eq-depth:

Age:	0..20	20..29	30-39	40-49	50-59	> 60
Tuples	1800	2000	2100	2200	1900	1800

Compressed: store separately some highly frequent values: (48,1900)

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Difficult Questions on Histograms

- Small number of buckets
 - Hundreds, or thousands, but not more
 - WHY ?
- Not updated during database update, but recomputed periodically
 - WHY ?
- Multidimensional histograms rarely used
 - WHY ?

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Summary of Query Optimization

- Three parts:
 - search space, algorithms, size/cost estimation
- Ideal goal: find optimal plan. But
 - Impossible to estimate accurately
 - Impossible to search the entire space
- Goal of today's optimizers:
 - Avoid very bad plans

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