CSE 444: Database Internals

Lecture 9
Query Plan Cost Estimation

Office Hour Announcement

- The day labs are due
  - No time to help debugging. Bring debugging problems to us before or after
  - We will schedule ad-hoc extra office hours. We will post on the message board

Before We Go Into Query Plan Costs… How do Updates Work? (Insert/Delete)

Example Using Delete

```
delete from R where a=1;
```

Query plan

Delete
|
Filter ($\sigma_{a=1}$)
|
SeqScan
|
R

In SimpleDB, the Delete Operator calls BufferPool.deleteTuple()

Why not call HeapFile.deleteTuple() directly?

Because there could also be indexes. Need some entity that will decide all the structures from where tuple needs to be deleted

BufferPool then calls HeapFile.deleteTuple()

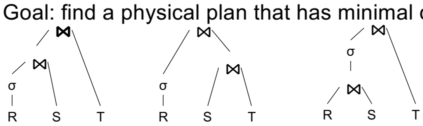
Pushing Updates to Disk

- When inserting a tuple, HeapFile inserts it on a page but does not write the page to disk
- When deleting a tuple, HeapFile deletes tuple from a page but does not write the page to disk
- The buffer manager worries when to write pages to disk (and when to read them from disk)
- When need to add new page to file, HeapFile adds page to file on disk and then reads it through buffer manager

Back to Query Optimization

Query Optimization Summary

Goal: find a physical plan that has minimal cost



What is the cost of a plan?

For each operator, cost is function of CPU, IO, network bw

Total_Cost = CPUCost + w_{IO} IOCost+ w_{BW} BWCost

Cost of plan is total for all operators

In this class, we look only at IO

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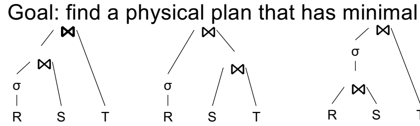

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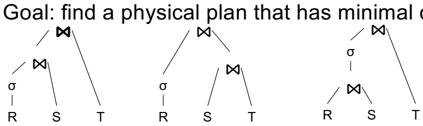

Know how to compute cost if know cardinalities

Query Optimization Summary

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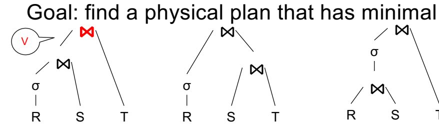

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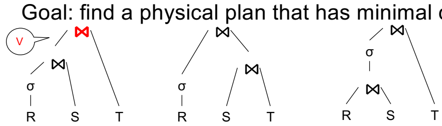

Know how to compute cost if know cardinalities

- Eg. Cost(V ⋈ T) = 3B(V) + 3B(T)
- B(V) = T(V) / PageSize
- T(V) = T(σ(R) ⋈ S)

Query Optimization Summary

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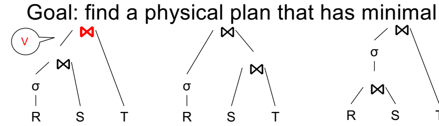

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- T(V) = T(σ(R) ⋈ S)

Cardinality estimation problem: e.g. estimate T(σ(R) ⋈ S)

Database Statistics

- Collect statistical summaries of stored data
- Estimate size (=cardinality) in a bottom-up fashion
  - This is the most difficult part, and still inadequate in today's query optimizers
- Estimate cost by using the estimated size
  - Hand-written formulas, similar to those we used for computing the cost of each physical operator

Database Statistics

- Number of tuples (cardinality) T(R)
- Indexes, number of keys in the index V(R,a)
- Number of physical pages B(R)
- Statistical information on attributes
  - Min value, Max value, V(R,a)
- Histograms

- Collection approach: periodic, using sampling

Size Estimation Problem

Q = SELECT list
FROM R1, …, Rn
WHERE $cond_1$ AND $cond_2$ AND . . . AND $cond_k$

Given T(R1), T(R2), …, T(Rn)
Estimate T(Q)

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

Size Estimation Problem

Q = SELECT list
FROM R1, …, Rn
WHERE $cond_1$ AND $cond_2$ AND . . . AND $cond_k$

Remark: T(Q) ≤ T(R1) × T(R2) × … × T(Rn)

Size Estimation Problem

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FROM R1, …, Rn
WHERE $cond_1$ AND $cond_2$ AND . . . AND $cond_k$

Remark: T(Q) ≤ T(R1) × T(R2) × … × T(Rn)

Key idea: each condition reduces the size of T(Q) by some factor, called selectivity factor

Selectivity Factor

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

Example

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

Q = 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

Q: What is the estimated size of the query output T(Q) ?

Example

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Q: What is the estimated size of the query output T(Q) ?

A: T(Q) = 30k * 200k * 10k * 1/3 * 1/10 * ½ = 10^{12}

Selectivity Factors for Conditions

- A = c /* $\sigma_{A=c}(R)$ */
  - Selectivity = 1/V(R,A)

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  - Selectivity = (c - Low(R, A))/(High(R,A) - Low(R,A))

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- A = c /* $\sigma_{A=c}(R)$ */
  - Selectivity = 1/V(R,A)
- A < c /* $\sigma_{A<c}(R)$*/
  - Selectivity = (c - Low(R, A))/(High(R,A) - Low(R,A))
- A = B /* $R \bowtie_{A=B} S$ */
  - Selectivity = 1 / max(V(R,A),V(S,A))
  - (will explain next)

Assumptions

- *Containment of values*: if V(R,A) <= V(S,B), then all values R.A occur in S.B
  - 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, $V(R \bowtie_{A=B} S, C) = V(R, C)$ (or V(S, C))
  - Note: we don't need this to estimate the size of the join, but we need it in estimating the next operator

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

Assume V(R,A) <= V(S,B)

- A 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)

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

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 ?

(In class…)

Complete Example

Supplier(<u>sid</u>, sname, scity, sstate)
Supply(<u>sid, pno</u>, quantity)

```
SELECT sname
FROM Supplier x, Supply y
WHERE x.sid = y.sid
  and y.pno = 2
  and x.scity = 'Seattle'
  and x.sstate = 'WA'
```

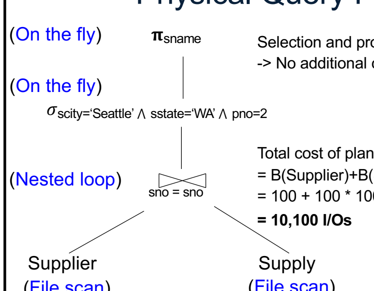
- Some statistics
  - T(Supplier) = 1000 records
  - T(Supply) = 10,000 records
  - B(Supplier) = 100 pages
  - B(Supply) = 100 pages
  - V(Supplier,scity) = 20, V(Suppliers,state) = 10
  - V(Supply,pno) = 2,500
  - Both relations are clustered
- M = 11

Computing the Cost of a Plan

- Estimate cardinality in a bottom-up fashion
  - Cardinality is the size of a relation (nb of tuples)
  - Compute size of *all* intermediate relations in plan
- Estimate cost by using the estimated cardinalities

T(Supplier) = 1000 B(Supplier) = 100 V(Supplier,scity) = 20 M = 11
T(Supply) = 10,000 B(Supply) = 100 V(Supplier,state) = 10
V(Supply,pno) = 2,500

Physical Query Plan 1



Selection and project on-the-fly
-> No additional cost.

Total cost of plan is thus cost of join:
= B(Supplier)+B(Supplier)*B(Supply)
= 100 + 100 * 100
= 10,100 I/Os

T(Supplier) = 1000 B(Supplier) = 100 V(Supplier,scity) = 20 M = 11
T(Supply) = 10,000 B(Supply) = 100 V(Supplier,state) = 10
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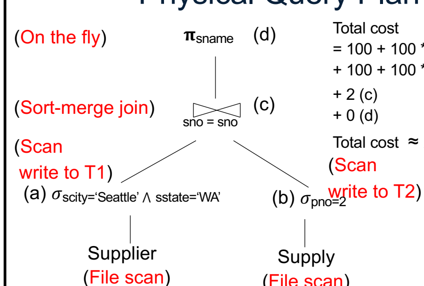
Physical Query Plan 2

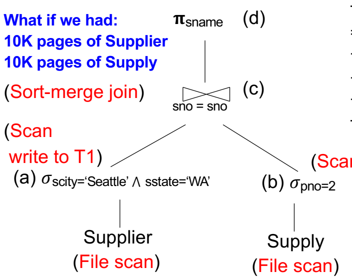


Total cost
= 100 + 100 * 1/20 * 1/10 (a)
+ 100 + 100 * 1/2500 (b)
+ 2 (c)
+ 0 (d)
Total cost ≈ **204 I/Os**

V(Supplier,scity) = 20 V(Supplier,state) = 10 V(Supply,pno) = 2,500 M = 11

Plan 2 with Different Numbers

What if we had:
10K pages of Supplier
10K pages of Supply

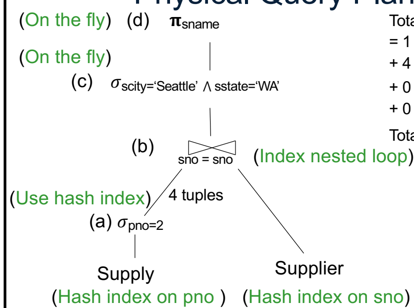


Total cost
= 10000 + 50 (a)
+ 10000 + 4 (b)
+ 3*50 + 4 (c)
+ 0 (d)
Total cost ≈ 20,208 I/Os

Need to do a two-pass sort algorithm

T(Supplier) = 1000 B(Supplier) = 100 V(Supplier,scity) = 20 M = 11
T(Supply) = 10,000 B(Supply) = 100 V(Supplier,state) = 10
V(Supply,pno) = 2,500

Physical Query Plan 3



Total cost
= 1 (a)
+ 4 (b)
+ 0 (c)
+ 0 (d)
Total cost ≈ **5 I/Os**

Assume: clustered
Clustering does not matter

Histograms

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

Histograms

Employee(<u>ssn</u>, name, age)

T(Employee) = 25000, V(Empolyee, age) = 50
min(age) = 19, max(age) = 68

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

Histograms

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Estimate = 25000 / 50 = 500 Estimate = 25000 * 6 / 50 = 3000

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

Histograms

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

Estimate = 1200 Estimate = 1*80 + 5*500 = 2580

Types of Histograms

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

Types of Histograms

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

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..33	33..38	38-43	43-45	45-54	> 54
Tuples	1800	2000	2100	2200	1900	1800

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

V-Optimal Histograms

- Defines bucket boundaries in an optimal way, to minimize the error over all point queries
- Computed rather expensively, using dynamic programming
- Modern databases systems use V-optimal histograms or some variations

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 ?

Difficult Questions on Histograms

- Small number of buckets
  - Hundreds, or thousands, but not more
  - WHY? All histograms are kept in main memory during query optimization; plus need fast access
- *Not* updated during database update, but recomputed periodically
  - WHY? Histogram update creates a write conflict; would dramatically slow down transaction throughput
- Multidimensional histograms rarely used
  - WHY? Too many possible multidimensional histograms, unclear which ones to choose