

CSE 444: Database Internals

Lecture 9 Query Plan Cost Estimation

CSE 444 - Winter 2018

1

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

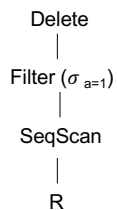
CSE 444 - Winter 2018

2

Example Using Delete

`delete from R where a=1;`

Query plan



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()`

CSE 444 - Winter 2018

3

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

CSE 444 - Winter 2018

4

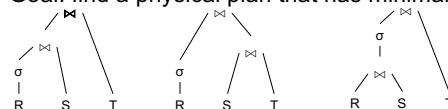
Back to Query Optimization

CSE 444 - Winter 2018

5

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

$$\text{Total_Cost} = \text{CPUCost} + w_{\text{IO}} \text{IOCost} + w_{\text{BW}} \text{BWCost}$$

Cost of plan is total for all operators

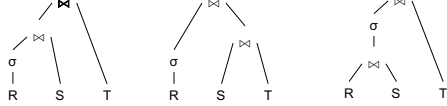
In this class, we look only at IO

CSE 444 - Winter 2018

6

Query Optimization Summary

Goal: find a physical plan that has minimal cost

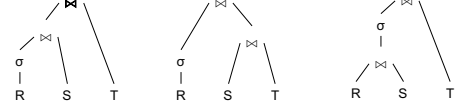


CSE 444 - Winter 2018

7

Query Optimization Summary

Goal: find a physical plan that has minimal cost



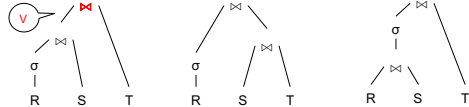
Know how to compute cost if know cardinalities

CSE 444 - Winter 2018

8

Query Optimization Summary

Goal: find a physical plan that has minimal cost



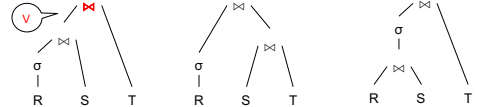
Know how to compute cost if know cardinalities

CSE 444 - Winter 2018

9

Query Optimization Summary

Goal: find a physical plan that has minimal cost



Know how to compute cost if know cardinalities

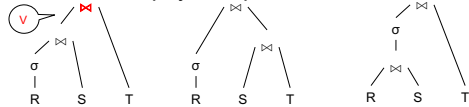
- Eg. $\text{Cost}(V \bowtie T) = 3B(V) + 3B(T)$
- $B(V) = T(V) / \text{PageSize}$
- $T(V) = T(\sigma(R) \bowtie S)$

CSE 444 - Winter 2018

10

Query Optimization Summary

Goal: find a physical plan that has minimal cost



Know how to compute cost if know cardinalities

- Eg. $\text{Cost}(V \bowtie T) = 3B(V) + 3B(T)$
- $B(V) = T(V) / \text{PageSize}$
- $T(V) = T(\sigma(R) \bowtie S)$

Cardinality estimation problem: e.g. estimate $T(\sigma(R) \bowtie S)$

CSE 444 - Winter 2018

11

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

CSE 444 - Winter 2018

12

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

CSE 444 - Winter 2018

13

Size Estimation Problem

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

Given $T(R_1), T(R_2), \dots, T(R_n)$
Estimate $T(Q)$

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

CSE 444 - Winter 2018

14

Size Estimation Problem

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

Remark: $T(Q) \leq T(R_1) \times T(R_2) \times \dots \times T(R_n)$

CSE 444 - Winter 2018

15

Size Estimation Problem

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

Remark: $T(Q) \leq T(R_1) \times T(R_2) \times \dots \times T(R_n)$

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

CSE 444 - Winter 2018

16

Selectivity Factor

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

CSE 444 - Winter 2018

17

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 $1/2$

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

CSE 444 - Winter 2018

18

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 1/2

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

A: T(Q) = 30k * 200k * 10k * 1/3 * 1/10 * 1/2 = 10¹²

Selectivity Factors for Conditions

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

CSE 444 - Winter 2018

20

Selectivity Factors for Conditions

- $A = c$ $/ * \sigma_{A=c}(R) * /$
– Selectivity = $1/V(R,A)$
- $A < c$ $/ * \sigma_{A<c}(R) * /$
– Selectivity = $(c - \text{Low}(R,A)) / (\text{High}(R,A) - \text{Low}(R,A))$

CSE 444 - Winter 2018

21

Selectivity Factors for Conditions

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

CSE 444 - Winter 2018

22

Assumptions

- **Containment of values:** if $V(R,A) \leq V(S,B)$, 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

CSE 444 - Winter 2018

23

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

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

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

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

CSE 444 - Winter 2018

24

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...)

CSE 444 - Winter 2018

25

Complete Example

Supplier(sid, sname, scity, sstate)
Supply(sid, pno, 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
 - $T(\text{Supplier}) = 1000$ records
 - $T(\text{Supply}) = 10,000$ records
 - $B(\text{Supplier}) = 100$ pages
 - $B(\text{Supply}) = 100$ pages
 - $V(\text{Supplier}, \text{scity}) = 20$, $V(\text{Suppliers}, \text{state}) = 10$
 - $V(\text{Supply}, \text{pno}) = 2,500$
 - Both relations are clustered
- $M = 11$

CSE 444 - Winter 2018

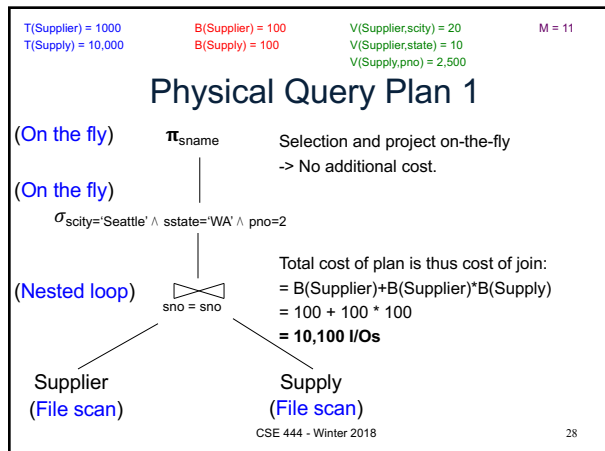
26

Computing the Cost of a Plan

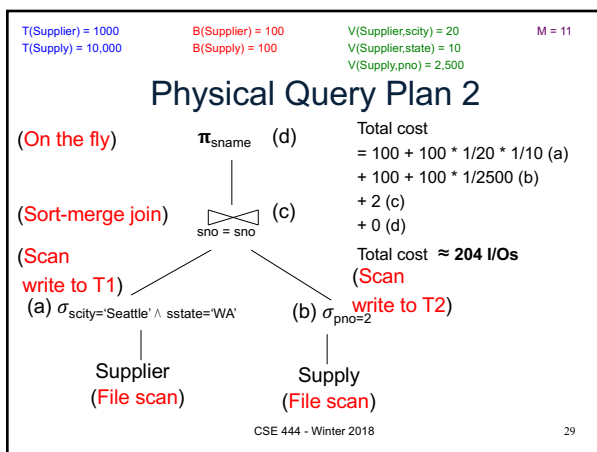
- Estimate cardinality in a bottom-up fashion
 - Cardinality is the size of a relation (nb of tuples)
 - Compute size of *all* intermediate relations in plan
- Estimate cost by using the estimated cardinalities

CSE 444 - Winter 2018

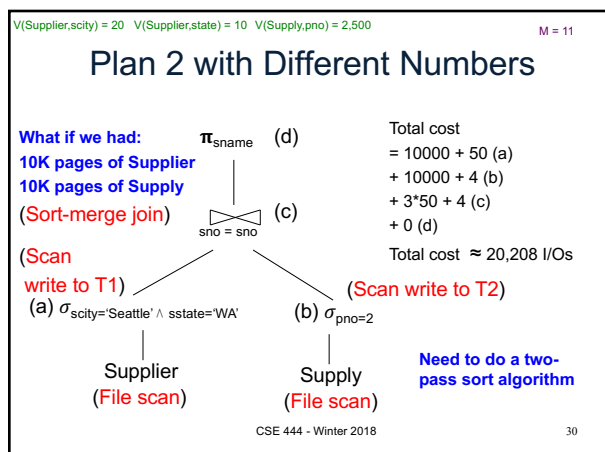
27



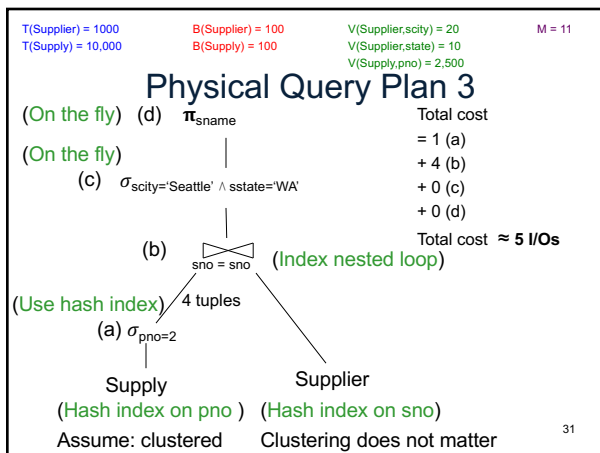
28



29



30



Histograms

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

CSE 444 - Winter 2018

32

Histograms

Employee(ssn, name, age)

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

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

CSE 444 - Winter 2018

33

Histograms

Employee(ssn, name, age)

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

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

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

CSE 444 - Winter 2018

34

Histograms

Employee(ssn, name, age)

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

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

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

CSE 444 - Winter 2018

35

Histograms

Employee(ssn, name, age)

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

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

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

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

36

Types of Histograms

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

CSE 444 - Winter 2018

37

Types of Histograms

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

CSE 444 - Winter 2018

38

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)

CSE 444 - Winter 2018

39

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

CSE 444 - Winter 2018

40

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 ?

CSE 444 - Winter 2018

41

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

CSE 444 - Winter 2018

42