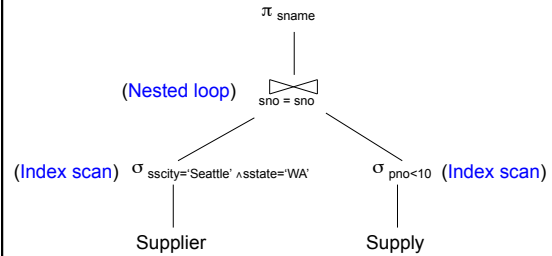


# CSE 444: Database Internals

## Lecture 9 Query Plan Cost Estimation

## Where We Are

Learning how to estimate cost of a query plan



## Where We Are

We already know how to

- Compute the cost of different operators
  - In terms of number IOs
  - Normally should also consider CPU and network
- Compute the cost of retrieving data from disk
  - Using a file scan
  - Using an index and an *equality* predicate

## Where We Are Going

We still need to

- Compute cost of retrieving data from disk
  - For more sophisticated predicates than equality
  - Compare the cost of different options
- Compute cost of a complete plan
  - How to put everything together

## Access Path

- **Access path**: a way to retrieve tuples from a table
  - A file scan
  - An index *plus* a matching selection condition
- Index *matches* selection condition if it can be used to retrieve just tuples that satisfy the condition
  - Example: `Supplier(sid, sname, scity, sstate)`
  - B+-tree index on `(sstate, scity)`
    - matches `sstate='WA'`
    - does not match `sid=3`, does not match `scity='Seattle'`

## Access Path Selection

- `Supplier(sid, sname, scity, sstate)`
- Selection condition: `sid > 300 ^ scity='Seattle'`
- Indexes: B+-tree on `sid` and B+-tree on `scity`
- Which access path should we use?
- We should pick the **most selective** access path

## Access Path Selectivity

- **Access path selectivity is the number of pages retrieved if we use this access path**
  - Most selective retrieves fewest pages
- As we saw earlier, **for equality predicates**
  - Selection on equality:  $\sigma_{a=v}(R)$
  - $V(R, a)$  = # of distinct values of attribute a
  - $1/V(R, a)$  is thus the reduction factor
  - Clustered index on a: cost  $B(R)/V(R, a)$
  - Unclustered index on a: cost  $T(R)/V(R, a)$
  - (we are ignoring I/O cost of index pages for simplicity)

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## Selectivity for Range Predicates

Selection on **range**:  $\sigma_{a>v}(R)$

- How to compute the selectivity?
- **Assume values are uniformly distributed**
- Reduction factor X
- $X = (\text{Max}(R, a) - v) / (\text{Max}(R, a) - \text{Min}(R, a))$
- Clustered index on a: cost  $B(R)*X$
- Unclustered index on a: cost  $T(R)*X$

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## Back to Our Example

- Selection condition: **sid > 300  $\wedge$  scity='Seattle'**
  - Index I1: B+-tree on sid clustered
  - Index I2: B+-tree on scity unclustered
- Let's assume
  - $V(\text{Supplier}, \text{scity}) = 20$
  - $\text{Max}(\text{Supplier}, \text{sid}) = 1000$ ,  $\text{Min}(\text{Supplier}, \text{sid}) = 1$
  - $B(\text{Supplier}) = 100$ ,  $T(\text{Supplier}) = 1000$
- **Cost I1:  $B(R) * (\text{Max}-v) / (\text{Max}-\text{Min}) = 100 * 700 / 999 \approx 70$**
- **Cost I2:  $T(R) * 1/V(\text{Supplier}, \text{scity}) = 1000 / 20 = 50$**

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## Selectivity with Multiple Conditions

What if we have an index on multiple attributes?

- Ex: selection  $\sigma_{a=v1 \wedge b=v2}(R)$  and index on  $\langle a, b \rangle$

How to compute the selectivity?

- **Assume attributes are independent**
- $X = 1 / (V(R, a) * V(R, b))$
- Clustered index on  $\langle a, b \rangle$ : cost  $B(R)*X$
- Unclustered index on  $\langle a, b \rangle$ : cost  $T(R)*X$

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## Access Path Selectivity Summary

First compute the reduction factor for predicate

- Predicate on single attribute
  - Equality:  $X = 1/V(R, a)$
  - Range:  $X = (\text{Max}(R, a) - v) / (\text{Max}(R, a) - \text{Min}(R, a))$
- Predicate on multiple attributes (using single index)
  - Compute reduction factor for each attribute  $X_1$  and  $X_2$
  - Multiply the reduction factors:  $X = X_1 * X_2$

Second compute the number of page IOs

- Clustered:  $B(R) * X$
- Unclustered  $T(R) * X$

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## Back to Estimating Cost of a Query Plan

- We already know how to
  - Compute the cost of different operations (last time)
  - Compute cost of retrieving tuples from disk with different access paths
- We still need to
  - **Compute cost of a complete plan**

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## 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
- We learned how to compute the cost last time
- But how do we compute cardinalities?

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## Projection: Cardinality of Result

- **Output cardinality same as input cardinality**
  - Same number of input tuples
  - But tuples are smaller!

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## Selection: Cardinality of Result

- **Multiply input cardinality by reduction factor**
  - Similar to estimating access path selectivity
  - In fact, we also say *operator/predicate selectivity*
  - Condition is  $a = c$  /\* value selection on R \*/
    - Selectivity =  $1/V(R,a)$
  - Condition is  $a < v$  /\* range selection on R \*/
    - Selectivity =  $(v - \text{Min}(R, a)) / (\text{Max}(R, a) - \text{Min}(R, a))$
  - Multiple conditions: assume independence
    - Use product of the reduction factors for the terms
    - Condition is  $a=v1 \wedge b=v2$
    - Selectivity =  $1 / (V(R,a) * V(R,b))$

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## Join: Cardinality of Result

- For **joins**  $R \bowtie S$ 
  - Take product of cardinalities of relations R and S
  - Apply reduction factors for each term in join condition
  - Terms are of the form: column1 = column2
  - Reduction:  $1 / (\text{MAX}(V(R, \text{column1}), V(S, \text{column2})))$ 
    - Assumes each value in smaller set has a matching value in the larger set (more on next slide)

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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,  $V(R \bowtie_{A=B} S, C) = V(R, C)$  (or  $V(S, C)$ )

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

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

- 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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## Complete Example

Supplier(sid, sname, scity, sstate)  
Supply(sid, pno, quantity)

- 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

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

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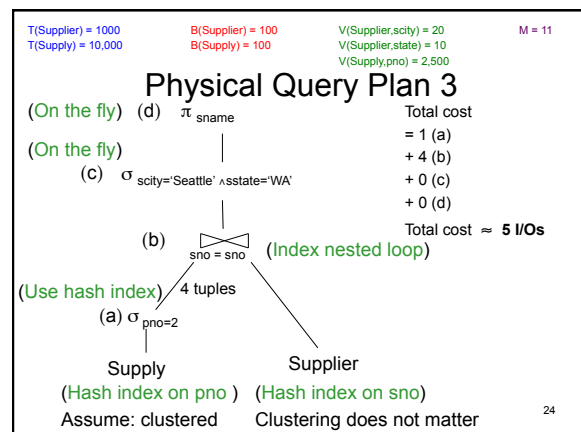
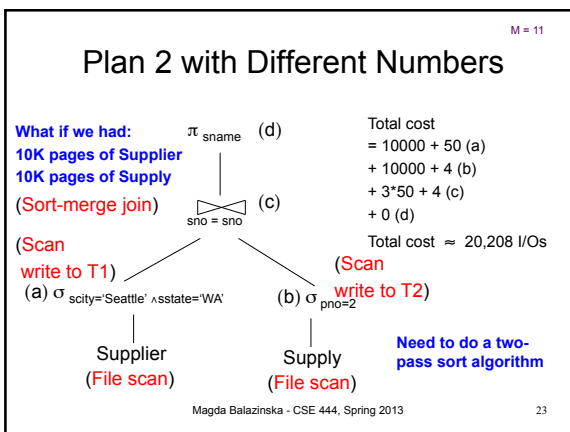
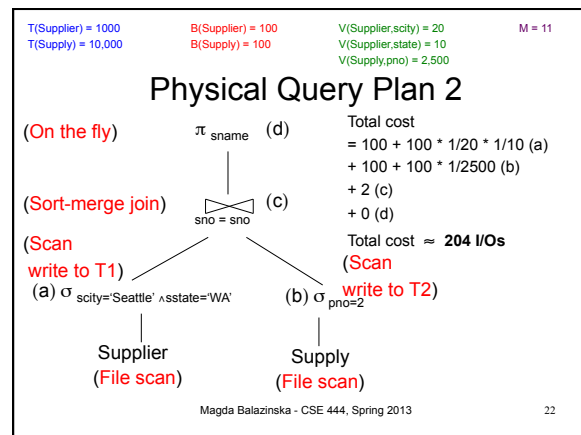
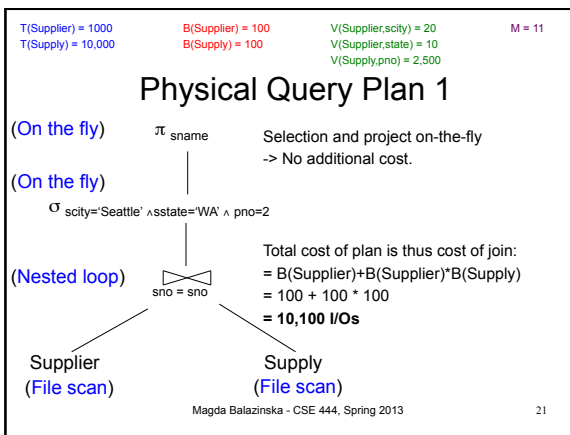
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## Computing the Cost of a Plan

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## Simplifications

- In the previous examples, we assumed that all index pages were in memory
- When this is not the case, we need to add the cost of fetching index pages from disk

## Different Cost Models

- In previous examples, we considered IO costs
- Typically, want IO+CPU
- For parallel/distributed queries, add network bw
- If need to compare *logical* plans
  - Compute the cardinality of each *intermediate* relation
  - Sum up all the cardinalities

## Statistics on Base Data

- All previous computations relied on information about the base relations
  - Number of blocks: B(R)
  - Number of tuples: T(R)
  - Min/max values
  - Number of distinct values: V(R,a)

## Statistics on Base Data

- **DBMSs collect a lot of info for base relations**
  - 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**

## Histograms

Employee(ssn, name, age)

T(Employee) = 25000, V(Employee, age) = 50  
min(age) = 8, max(age) = 68

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

## Histograms

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Estimate = 25000 / 50 = 500



Estimate = 25000 \* 6 / 60 = 2500

## Histograms

**Employee(ssn, name, age)**

$T(\text{Employee}) = 25000$ ,  $V(\text{Employee, age}) = 50$   
 $\min(\text{age}) = 8$ ,  $\max(\text{age}) = 68$

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

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

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## Histograms

**Employee(ssn, name, age)**

$T(\text{Employee}) = 25000$ ,  $V(\text{Employee, age}) = 50$   
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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

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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..35 | 36..40 | 41-44 | 45-52 | 53-56 | > 57 |
| Tuples | 1800  | 2000   | 2100  | 2200  | 1900  | 1800 |

**Compressed:** store separately 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

- What we know
  - Different types of physical query plans
  - How to compute the cost of a query plan
  - Although it is hard to compute the cost accurately
- We can now compare query plans!
- Next: Query optimization