

Relational Algebra

- Operates on relations, i.e. *sets* - Later: we discuss how to extend this to *bags*
- Later: we discuss how to extend Five operators:
- Union: ∪
- Difference: -
- Selection: σ
- Projection: Π
 Cartesian Product: ×
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 Derived or auxiliary operators:
 - Intersection, complement
 - Joins (natural, equi-join, theta join, semi-join)
 - Renaming: ρ

1. Union and 2. Difference

- $R_1 \cup R_2$
- Example: ActiveEmployees U RetiredEmployees
- $R_1 R_2$
- Example: AllEmployees – RetiredEmployees

What about Intersection ?

- It is a derived operator
- $R_1 \cap R_2 = R_1 (R_1 R_2)$
- Also expressed as a join (will see later)
- Example UnionizedEmployees ∩ RetiredEmployees

3. Selection

- Returns all tuples which satisfy a condition
- Notation: $\sigma_c(R)$
- Examples
 - $\sigma_{_{Salary > 40000}}$ (Employee)
 - $\sigma_{name = "Smithh"}$ (Employee)
- The condition c can be =, <, \leq , >, \geq , <>

4. Projection

- · Eliminates columns, then removes duplicates
- Notation: $\Pi_{A1,...,An}(R)$
- Example: project social-security number and names:
 - Π_{SSN, Name} (Employee) Output schema: Answer(SSN, Name)

5. Cartesian Product

- Each tuple in R_1 with each tuple in R_2
- Notation: $R_1 \times R_2$
- Example:
 - Employee × Dependents
- Very rare in practice; mainly used to express joins

Cartesian Product Example

SSN	
999999999	
77777777	
	999999999

Dependents	
EmployeeSSN	Dname
9999999999	Emily
777777777	Joe

Employee x Dependents				
Name	SSN	EmployeeSSN	Dname	
John	9999999999	9999999999	Emily	
John	9999999999	777777777	Joe	
Tony	777777777	9999999999	Emily	
Tony	777777777	777777777	Joe	

Renaming

- Changes the schema, not the instance
- Notation: $\rho_{B1,...,Bn}(R)$
- Example:

ρ_{LastName, SocSocNo} (Employee) Output schema: Answer(LastName, SocSocNo)

Renaming Example

Employee

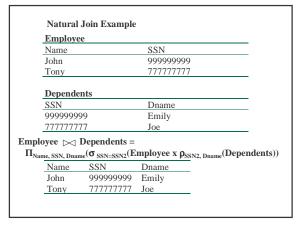
Name	SSN
John	999999999
Tony	77777777

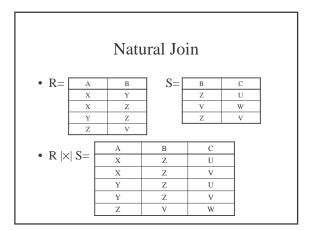
$\rho_{\text{LastName, SocSocNo}} \left(\text{Employee} \right)$

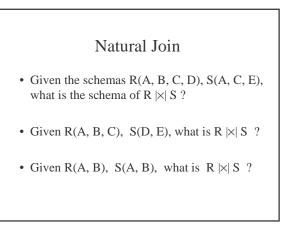
LastName	SocSocNo	
John	999999999	
Tony	77777777	

Natural Join

- Notation: $\mathbf{R}_1 \mid \times \mid \mathbf{R}_2$
- Meaning: $R_1 |\times| R_2 = \Pi_A(\sigma_C(R_1 \times R_2))$
- Where:
 - The selection $\sigma_{\!C}\,checks$ equality of all common attributes
 - The projection eliminates the duplicate common attributes







Theta Join

- A join that involves a predicate
- R1 $\mid \times \mid_{\theta}$ R2 = σ_{θ} (R1 × R2)
- Here θ can be any condition: =, <, \neq , \leq , \geq

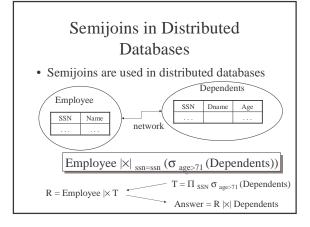
Eq-join

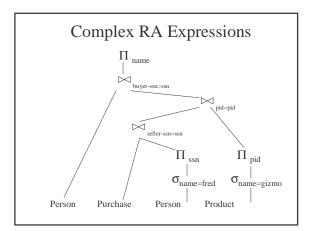
- A theta join where θ is an equality
- $R_1 \mid \times \mid_{A=B} R_2 = \sigma_{A=B} (R_1 \times R_2)$
- Example: Employee |×| _{SSN=SSN} Dependents
- Most useful join in practice

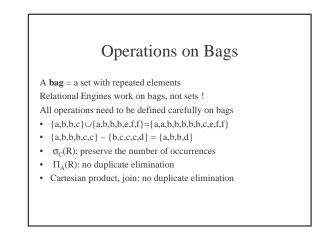
Semijoin

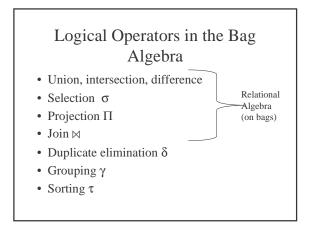
- $R \mid \times S = \prod_{A1,...,An} (R \mid \times \mid S)$
- Where $A_1, ..., A_n$ are the attributes in R
- Example:

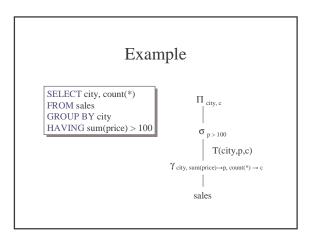
Employee |× Dependents

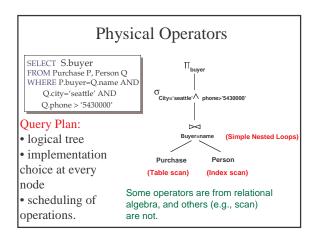


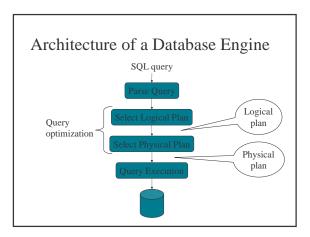












Cost Parameters

In database systems the data is on disks, not in main memory

The *cost* of an operation = total number of I/Os Cost parameters:

- B(R) = number of blocks for relation R
- T(R) = number of tuples in relation R
- V(R, a) = number of distinct values of attribute a

Cost Parameters

- Clustered table R:
 - Blocks consists only of records from this table
 - $B(R) \approx T(R) / blockSize$
- Unclustered table R:
 - Its records are placed on blocks with other tables
 When R is *unclustered*: B(R) ≈ T(R)
- When a is a key, V(R,a) = T(R)
- When a is not a key, V(R,a)

Cost

Cost of an operation = number of disk I/Os needed to:

read the operands

- compute the result

Cost of writing the result to disk is *not included* on the following slides

<u>Question</u>: the cost of sorting a table with B blocks ? <u>Answer</u>:

Scanning Tables

- The table is *clustered*:
 - Table-scan: if we know where the blocks are
 - Index scan: if we have a sparse index to find the blocks
- The table is *unclustered*
 - May need one read for each record

Sorting While Scanning

• Sometimes it is useful to have the output sorted

- Three ways to scan it sorted:
 - If there is a primary or secondary index on it, use it during scan
 - If it fits in memory, sort there
 - If not, use multi-way merge sort

Cost of the Scan Operator

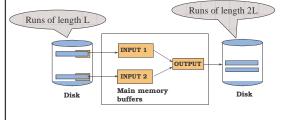
- Clustered relation:
 - Table scan:
 - Unsorted: B(R)Sorted: 3B(R)
 - Index scan
 - Unsorted: B(R)
 - Sorted: B(R) or 3B(R)
- Unclustered relation
 - Unsorted: T(R)
 - Sorted: T(R) + 2B(R)

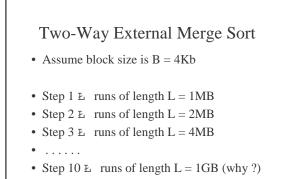
Sorting

- Problem: sort 1 GB of data with 1MB of RAM.
- Where we need this:
 - Data requested in sorted order (ORDER BY)
 - Needed for grouping operations
 - First step in sort-merge join algorithm
 - Duplicate removal
 - Bulk loading of B+-tree indexes.

2-Way Merge-sort: Requires 3 Buffers in RAM

- Pass 1: Read 1MB, sort it, write it.
- Pass 2, 3, ..., etc.: merge two runs, write them





Need 10 iterations over the disk data to sort 1GB

Can We Do Better ?

• Hint:

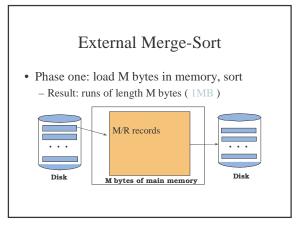
We have 1MB of main memory, but only used 12KB

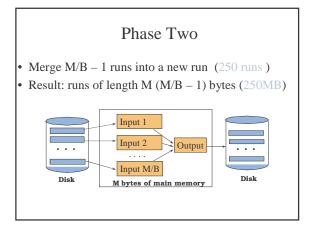
Cost Model for Our Analysis

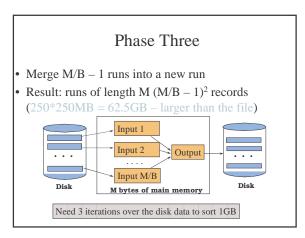
- **B:** Block size (= 4KB)
- M: Size of main memory (= 1MB)

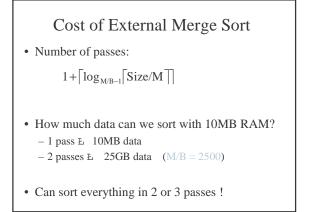
For later use (won't need now):

- N: Number of records in the file
- R: Size of one record











- The **xsort** tool in the XML toolkit sorts using this algorithm
- Can sort 1GB of XML data in about 8 minutes