

CSE 444: Database Internals

Lecture 20 Distributed Query Processing and Optimization

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References

- In our book: Sections 20.3 and 20.4
- Other book: **Database management systems.** Ramakrishnan and Gehrke. Third Ed. Chapter 22

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Where We Are

- We know all about a single-machine DBMS
- What if we had more than one machine?
 - Distributed query processing (today)
 - Distributed transactions (next lecture)
 - Replication (next week)
 - Parallel query processing (later next week)

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Outline

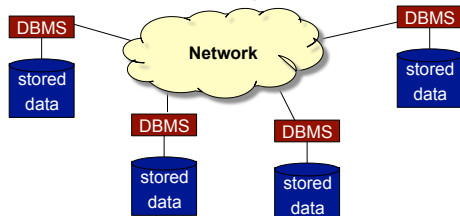
- Distributed DBMS motivation
- Distributed query optimization/processing
- Distributed DBMS limitations and challenges

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Distributed DBMS

- Important: many forms and definitions
- Our definition: shared nothing infrastructure
 - Multiple machines connected with a network
 - Machines can be far away from each other



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Vertical Partitioning

Students

StudentID	Name	Address	GPA	Status
234234	Mary	Houston	3.5	Scholarship
345345	Sue	Seattle	2.9	Fellowship
345343	Joan	Seattle	3.2	Veteran
234234	Ann	Portland	3.9	None



Registrar

StudentID	Name	Address	GPA
234234	Mary	Houston	3.5
345345	Sue	Seattle	2.9
...

Bursar

SID	Status
234234	Scholarship
345345	Fellowship
...	...

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Horizontal Partitioning

Customers

SSN	Name	City	Country
234234	Mary	Houston	USA
345345	Sue	Seattle	USA
345343	Joan	Seattle	USA
234234	Ann	Portland	USA
--	Frank	Calgary	Canada
--	Jean	Montreal	Canada



CustomersInHouston

SSN	Name	City	Country
234234	Mary	Houston	USA

CustomersInSeattle

SSN	Name	City	Country
345345	Sue	Seattle	USA
345343	Joan	Seattle	USA

CustomersInCanada

SSN	Name	City	Country
--	Frank	Calgary	Canada
--	Jean	Montreal	Canada

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Reasons for a Distributed DBMS

- **Scalability** (ex: Amazon, eBay, Google)
 - Many small servers cheaper than one big server
 - Need to scale incrementally
 - Typically: horizontal partitioning
- **Inherent distribution**
 - Large organizations have data at multiple locations (different offices) -> original motivation
 - Different types of data in different DBMSs
 - Web-based and Internet-based applications
 - Either horizontal or vertical partitioning

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Distributed vs Parallel DBMS

- **Parallel DBMS**
 - Goal to improve performance through parallelization
 - Parallel load, parallel index build, parallel query processing
 - Data distribution is governed solely by performance
 - Typically all machines are on the same cluster
- **Distributed DBMS**
 - Data stored across several sites
 - Each site capable of running independent of other sites
 - Each site can be administered autonomously
 - Data distribution not only based on performance

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Goals of a Distributed DBMS

- **Shield users from distribution details**
- **Distribution transparency**
 - **Naming** transparency
 - **Location** transparency
 - **Fragmentation** transparency
 - **Performance** transparency
 - Distributed query optimizer ensures similar performance no matter where query is submitted
 - **Schema change** and **transaction** transparency
- **Replication** transparency
- and more...

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Distributed DBMS Features

- 70's and 80's, three main prototypes:
 - SDD-1, distributed INGRES, and R*
- Main components of a distributed DBMS
 - Defining **data placement and fragmentation**
 - Distributed **catalog**
 - Distributed **query optimization** (today)
 - Distributed **transactions** (next lecture)
 - Managing **replicas** (next week)

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Outline

- Distributed DBMS motivation
- **Distributed query optimization/processing**
- Distributed DBMS limitations and challenges

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Review: Query Evaluation

Steps of query evaluation

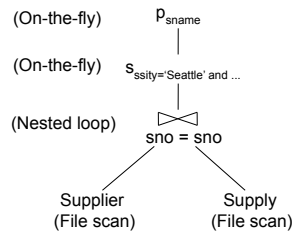
Query Parser

Query Rewrite

Optimizer

Executor

Query plan



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Review: Query Optimization

- Enumerate alternative plans
 - Many possible equivalent trees: e.g., join order
 - Many implementations for each operator
- Compute estimated cost of each plan
 - Compute number of I/Os and CPU utilization
 - Based on statistics
- Choose plan with lowest cost

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Distributed Query Optimization

- Search space is larger
 - Must **select sites** for joining relations
 - Must select **method for shipping inner relation**: whole or matches
- Minimize resource utilization
 - I/O, CPU, & **communication costs**
 - Example cost function used in R*
 - $W_{CPU} Nb_{inst} + W_{I/O} Nb_{I/O} + W_{msg} Nb_{msg} + W_{byte} Nb_{bytes}$
- Could also try to minimize response time
 - Least cost plan != Fastest plan

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Inner Table Transfer Strategy

- Ship whole
 - Read inner relation at its home site (using index or not)
 - Project inner relation to remove attributes not needed
 - Apply any single-table predicates
 - Ship results to site of outer relation and store in temp file
 - Note: we lose any indexes on the inner relation
- Fetch matches
 - For each tuple of outer, project tuple on join column
 - Send value to site of inner relation
 - Find matching tuples from inner relation
 - Ship projected, matching tuples back

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Additional Join Strategies

- Dynamically-created temporary index on inner
 - Ship inner relation, store in temp table, index temp
- Semijoin
 - Project outer relation on join column (eliminate dups)
 - Ship projected column to site with inner relation
 - Compute natural join and ship matching tuples back
 - Join the two relations
- Bloomjoin
 - Same idea as semijoin, but use Bloom filter instead of sending all values in the join column
 - Bloom filter creates some false positives through collisions

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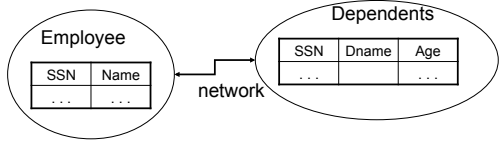
Semijoin

- $R \bowtie S = \Pi_{A_1, \dots, A_n} (R \bowtie S)$
- Where A_1, \dots, A_n are the attributes in R
- Example:
 - Employee \bowtie Dependents

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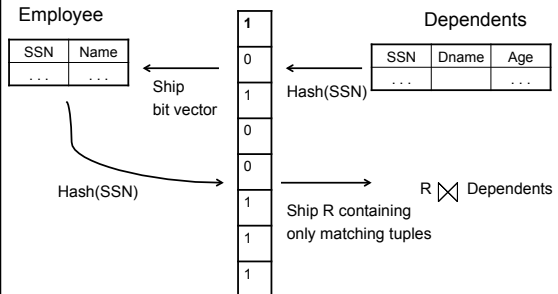
Semijoins in Distributed Databases



$\text{Employee} \bowtie_{\text{SSN}=\text{SSN}} (\sigma_{\text{age}>71}(\text{Dependents}))$

$R = \text{Employee} \bowtie T$
 $T = \Pi_{\text{SSN}} (\sigma_{\text{age}>71}(\text{Dependents}))$
 Answer = $R \bowtie \text{Dependents}$

Bloomjoin



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Distributed vs Local Joins

Why can distributed joins be faster than local ones?

- More resources are available to the join
 - Ex: Distributed query can use twice the buffer pool (useful when accessing relations through unclustered indexes)
- Different parts of the join can proceed in parallel
 - Ex: Join tuples from page 1 while shipping page 2
 - Ex: Can sort the two relations in parallel

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Outline

- Distributed DBMS motivation
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Distributed DBMS Limitations

- Top-down
 - Global, a priori data placement
 - Global query optimization, one query at a time; no notion of load balance
 - Distributed transactions, tight coupling
- Assumes full cooperation of all sites
- Assumes uniform sites
- Assumes short-duration operations
- Limited scalability

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Distributed DBMS Challenges

In some distributed databases

- **Autonomy**: different administrative domains
 - Cannot always assume full cooperation
 - Do not want distributed transactions
- **Heterogeneity**
 - Different capabilities at different locations
 - Different data types, different semantics -> data integration pb
- **Large-scale**
 - Internet-scale query processor

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