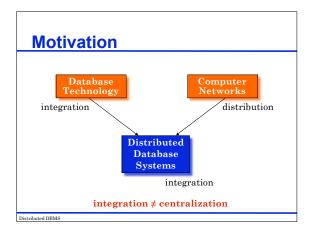
Distributed Database Management Systems

Outline

- Introduction
- Distributed DBMS Architecture
- Distributed Database Design
- Distributed Query Processing
- Distributed Concurrency Control
- Distributed Reliability Protocols

Distributed DBMS

Outline Introduction What is a distributed DBMS Problems Current state-of-affairs Distributed DBMS Architecture Distributed Database Design Distributed Query Processing Distributed Concurrency Control Distributed Reliability Protocols

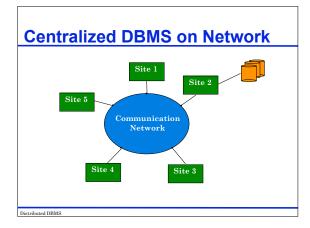




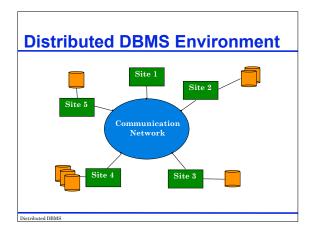
What is a Distributed Database System?

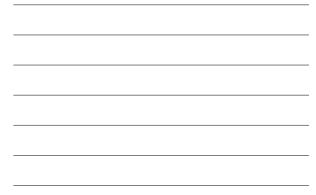
- A distributed database (DDB) is a collection of multiple, logically interrelated databases distributed over a computer network.
- A distributed database management system (D–DBMS) is the software that manages the DDB and provides an access mechanism that makes this distribution transparent to the users.

Distributed database system (DDBS) = DDB + D–DBMS









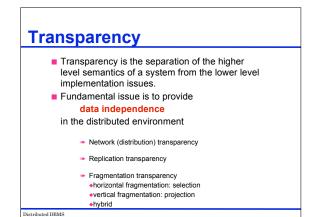
Implicit Assumptions

- Data stored at a number of sites ⇒ each site logically consists of a single processor.
- Processors at different sites are interconnected by a computer network ⇔ no multiprocessors → parallel database systems
- Distributed database is a database, not a collection of files
 data logically related as exhibited in the users' access patterns
 relational data model
- D-DBMS is a full-fledged DBMS
 - not remote file system, not a TP system

Distributed DBMS

Distributed DBMS Promises

- Transparent management of distributed, fragmented, and replicated data
- Improved reliability/availability through distributed transactions
- Improved performance
- **4** Easier and more economical system expansion

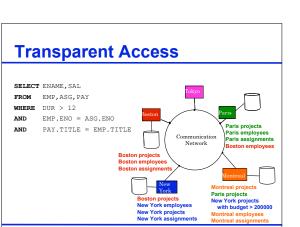


Example EMP ASG ENO ENAME DUR TITLE eno pno RESP J. Doe M. Smith A. Lee J. Miller B. Casey L. Chu R. Davis Elect. Eng. Syst. Anal. Mech. Eng. Programme Syst. Anal. Elect. Eng. Mech. Eng. Manager Analyst Consultant Engineer Programmer Manager Manager Engineer Engineer E1 E2 E3 E4 E5 E6 E7 E1 E2 E3 E3 E4 E5 E6 E7 E7 E8 P1 P2 P3 P4 P2 P2 P4 P3 P3 P5 $\frac{12}{24}$ $\begin{array}{r}
 6 \\
 10 \\
 48 \\
 24 \\
 48 \\
 36 \\
 23
 \end{array}$ E8 J. Jones st. Anal. P3 Manager 40 PROJ PAY BUDGET PNO PNAME TITLE SAL
 P1
 Instrumentation

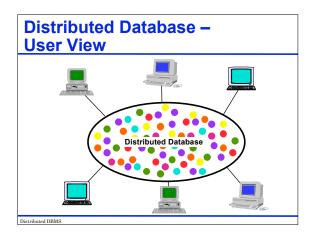
 P2
 Database Develop.

 P3
 CAD/CAM

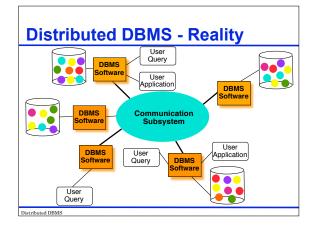
 P4
 Maintenance
 Elect. Eng. Syst. Anal. Mech. Eng. Programme: 40000 34000 27000 24000 $150000 \\ 135000$ $250000 \\ 310000$ Distributed DBMS



4









Potentially Improved Performance

- Proximity of data to its points of use
 - Requires some support for fragmentation and replication
- Parallelism in execution
 - Inter-query parallelism
 - Intra-query parallelism



System Expansion

- Issue is database scaling
- Emergence of microprocessor and workstation technologies
 - Demise of Grosh's law
 - Client-server model of computing
- Data communication cost vs telecommunication cost

Distributed DBMS

Distributed DBMS

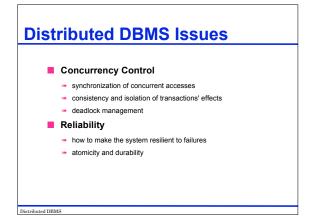
Distributed DBMS Issues

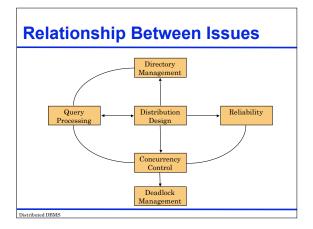
Distributed Database Design

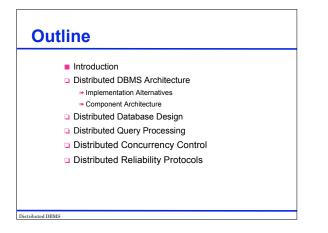
- how to distribute the database
- replicated & non-replicated database distribution
- a related problem in directory management

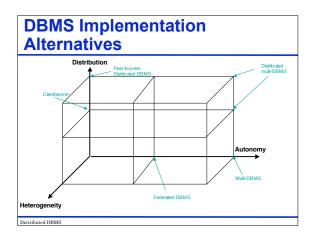
Query Processing

- convert user transactions to data manipulation instructions
- optimization problem
- min{cost = data transmission + local processing}
- general formulation is NP-hard







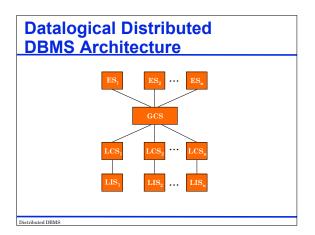


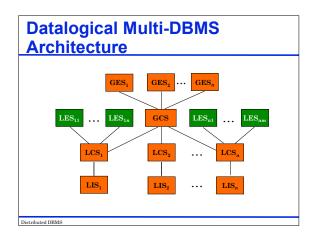


Dimensions of the Problem

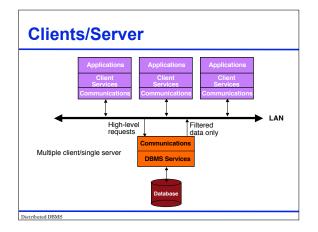
- Distribution
 - Whether the components of the system are located on the same machine or not
- Heterogeneity
 - Various levels (hardware, communications, operating system) DBMS important one
- data model, query language,transaction management algorithms Autonomy
 - Not well understood and most troublesome

 - Various versions
 Design autonomy: Ability of a component DBMS to decide on Design autonomy: Ability or a component DBMs to decide on issues related to its own design.
 Communication autonomy: Ability of a component DBMS to decide whether and how to communicate with other DBMSs.
 Execution autonomy: Ability of a component DBMS to execute local operations in any manner it wants to.

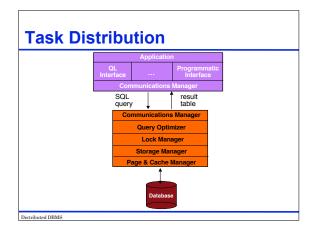


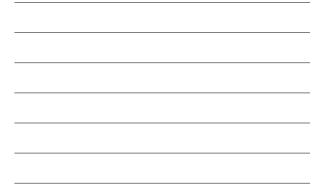










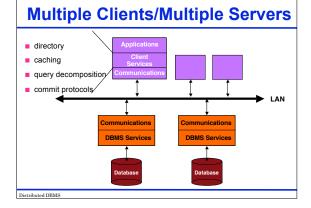


Advantages of Client-Server Architectures

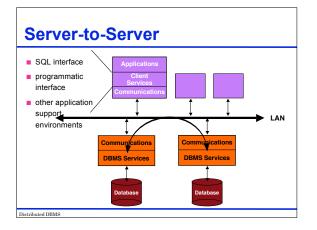
- More efficient division of labor
- Horizontal and vertical scaling of resources
- Better price/performance on client machines
- Ability to use familiar tools on client machines
- Client access to remote data (via standards)
- Full DBMS functionality provided to client workstations
- Overall better system price/performance

Distributed DBMS

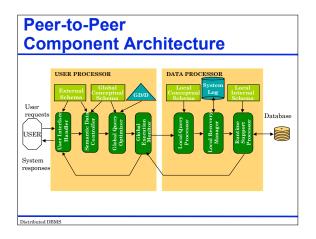
Problems With Multiple-Client/Single Server • Server forms bottleneck • Server forms single point of failure • Database scaling difficult

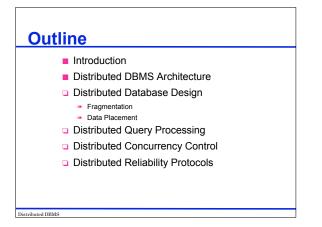


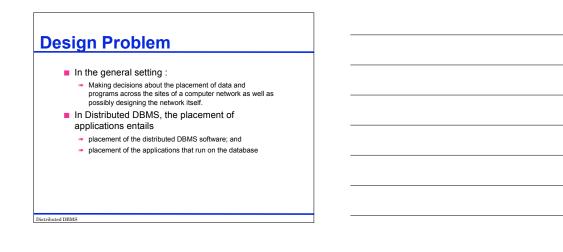


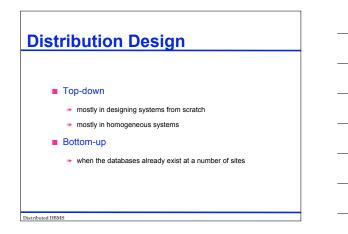


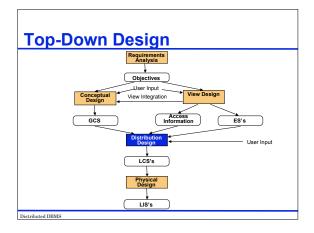




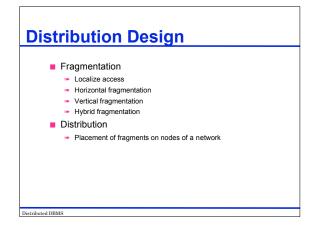






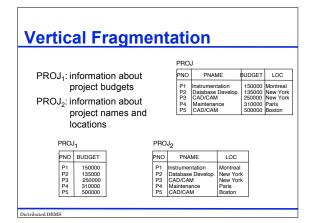






Horizontal Fragmentation PROJ PROJ_1 : projects with budgets PNO PNAME BUDGET LOC P1 Instrumentation P2 Database Devel P3 CAD/CAM P4 Maintenance P5 CAD/CAM 150000 Montreal 135000 New York 250000 New York 310000 Paris 500000 Boston less than \$200,000 PROJ₂ : projects with budgets greater than or equal to \$200,000 PROJ₂ PROJ₁ BUDGET LOC PNAME BUDGET LOC PNO PNAME PNO 50000 P3 CAD/CAM 250000 P1 Montreal New York entation strun P4 310000 P2 atabase Develop. 135000 New York Maintenance Paris P5 CAD/CAM 500000 Boston Distributed DBMS







Correctness of Fragmentation

- Completeness
 - Decomposition of relation R into fragments R₁, R₂, ..., R_n is complete iff each data item in R can also be found in some R_i
- Reconstruction
 - If relation R is decomposed into fragments R₁, R₂, ..., R_n, then there should exist some relational operator ⊽ such that
 R = ∇_{1≤j≤n}R_j

Disjointness

If relation R is decomposed into fragments R₁, R₂, ..., R₂, and data item d₁ is in R₂, then d₁ should not be in any other fragment R₄ (k ≠ j).

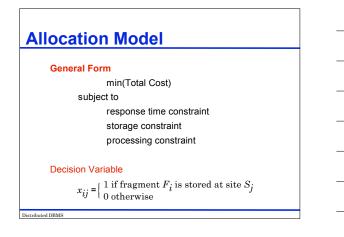
Distributed DBMS

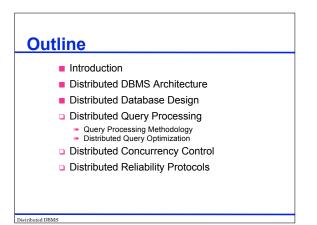
Allocation Alternatives

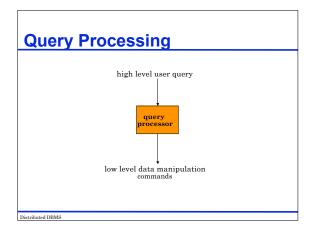
- Non-replicated
- partitioned : each fragment resides at only one site
 Replicated
 - fully replicated : each fragment at each site
- partially replicated : each fragment at some of the sites
 Rule of thumb:

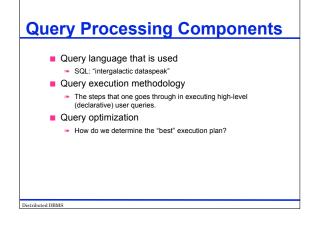
If read - only queries > 1 replication is advantageous, otherwise replication may cause problems

Frag	gment Allocation
	Problem Statement Given • F = {F1, F2,, Fn} fragments • S ={S1, S2,, Sm} network sites
	 Q = {q1, q2,, qq} applications Find the "optimal" distribution of F to S. Optimality Minimal cost
	 Communication + storage + processing (read & update) Cost in terms of time (usually) Performance
	 Response time and/or throughput Constraints Per site constraints (storage & processing)

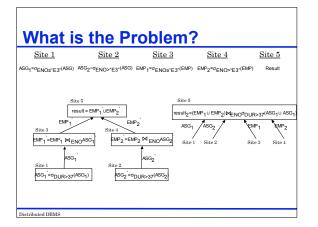






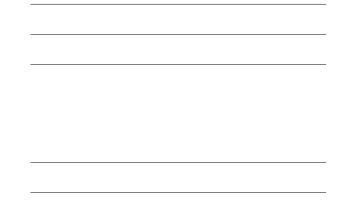


Selecting Alternatives





ost o	f Alternatives	
Assum	e:	
-	size(EMP) = 400, size(ASG) = 1000	
-	tuple access cost = 1 unit; tuple transfer cost = 10 units	
Strateg	y 1	
0	produce ASG': (10+10)*tuple access cost	20
0	ransfer ASG' to the sites of EMP: (10+10)*tuple transfer cost	200
8	produce EMP': (10+10) *tuple access cost*2	40
0	ransfer EMP' to result site: (10+10) *tuple transfer cost	200
	Total cost	460
Strateg	y 2	
0	ransfer EMP to site 5:400∗tuple transfer cost	4,000
0	ransfer ASG to site 5 :1000∗tuple transfer cost	10,000
8	produce ASG':1000*tuple access cost	1,000
0	oin EMP and ASG':400*20*tuple access cost	8.000
	Total cost	23,000



Query Optimization Objectives
Minimize a cost function
I/O cost + CPU cost + communication cost
These might have different weights in different distributed environments
Wide area networks
 communication cost will dominate
 low bandwidth
 low speed
 high protocol overhead
 most algorithms ignore all other cost components
Local area networks
 communication cost not that dominant
 total cost function should be considered
Can also maximize throughput
Distributed DBMS

Query Optimization Issues – Types of Optimizers

- Exhaustive search
 - cost-based
 - optimal
 - combinatorial complexity in the number of relations
- Heuristics
 - not optimal
 - regroup common sub-expressions
 - perform selection, projection first

 - replace a join by a series of semijoins
 reorder operations to reduce intermediate relation size
 optimize individual operations

Query Optimization Issues – Optimization Granularity Single query at a time cannot use common intermediate results Multiple queries at a time efficient if many similar queries decision space is much larger

Query Optimization Issues – Optimization Timing

Static

Distributed DBMS

- compilation ⇒ optimize prior to the execution
 difficult to estimate the size of the intermediate results ⇒ error propagation
- can amortize over many executions
 R*

Dynamic

- run time optimization
- exact information on the intermediate relation sizes
- have to reoptimize for multiple executions
- Distributed INGRES
- Hybrid
 - compile using a static algorithm
 - if the error in estimate sizes > threshold, reoptimize at run time

MERMAID

Distributed DBMS

Query Optimization Issues – Statistics

Relation

 cardinality
 size of a tuple
 fraction of tuples participating in a join with another relation

 Attribute

 cardinality of domain

actual number of distinct values

Common assumptions

- independence between different attribute values
- uniform distribution of attribute values within their domain

Query Optimization Issues – Decision Sites

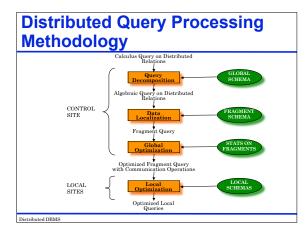
Centralized

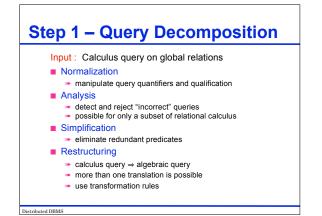
- single site determines the "best" schedule
- simple
- need knowledge about the entire distributed database
- Distributed
 - · cooperation among sites to determine the schedule
 - need only local information
 - cost of cooperation
- Hybrid
 - one site determines the global schedule
 - each site optimizes the local subqueries

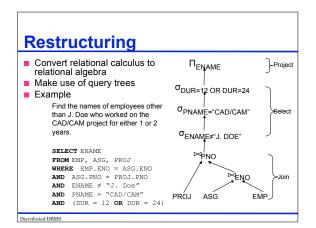
Distributed DBMS

Query Optimization Issues – Network Topology Wide area networks (WAN) – point-to-point • characteristics

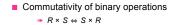
- characteristics
- low bandwidth low speed
- high protocol overhead
- communication cost will dominate; ignore all other cost
- factors
- global schedule to minimize communication cost
- local schedules according to centralized query optimization
- Local area networks (LAN)
 - communication cost not that dominant
 total cost function should be considered
 - broadcasting can be exploited (joins)
 - special algorithms exist for star networks



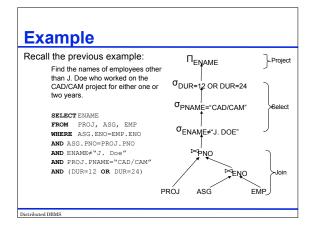




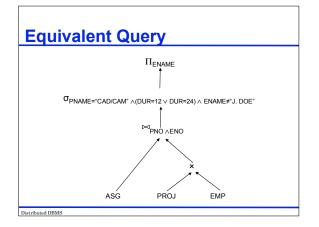
Restructuring – Transformation Rules (Examples)



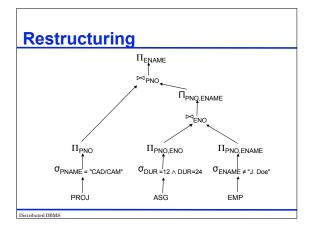
- R ⋈S ⇔ S ⋈R
- R ∪ S ⇔ S ∪ R
- Associativity of binary operations
 - $\twoheadrightarrow (R \times S) \times T \Leftrightarrow R \times (S \times T)$
 - $(R \bowtie S) \bowtie T \Leftrightarrow R \bowtie (S \bowtie T)$
- Idempotence of unary operations
 - → Π_A['](Π_A['](R)) ⇔ Π_A['](R)
 - $\sigma_{p_1(A_1)}(\sigma_{p_2(A_2)}(R)) = \sigma_{p_1(A_1)} \wedge_{p_2(A_2)}(R)$ where R[A] and $A' \subseteq A$, $A'' \subseteq A$ and $A' \subseteq A''$
- Commuting selection with projection



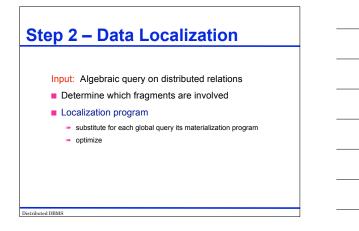


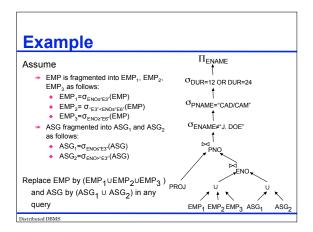




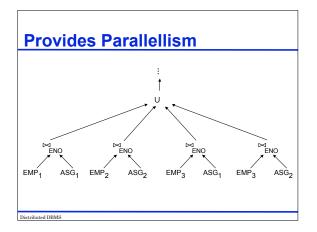




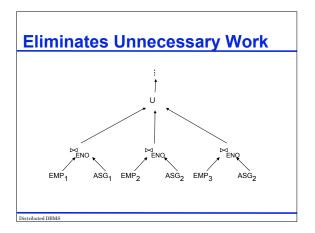


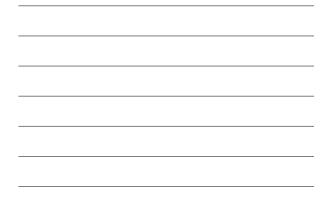










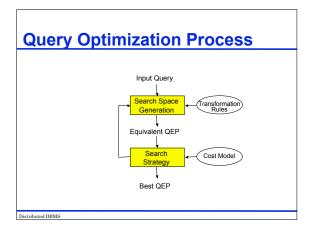


Step 3 – Global Query Optimization Input: Fragment query Find the *best* (not necessarily optimal) global schedule Minimize a cost function Distributed join processing Bushy vs. linear trees Which relation to ship where? Ship-whole vs ship-as-needed Decide on the use of semijoins Semijoin saves on communication at the expense of more local processing. Join methods nested loop vs ordered joins (merge join or hash join)

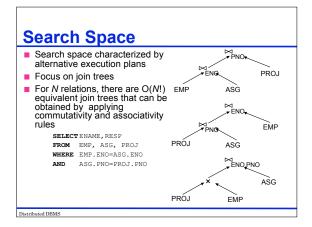
Cost-Based Optimization

Solution space

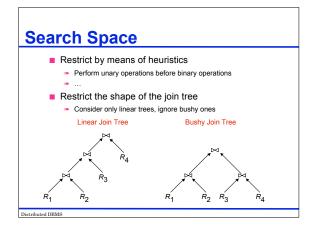
- The set of equivalent algebra expressions (query trees).
- Cost function (in terms of time)
 - I/O cost + CPU cost + communication cost
 - These might have different weights in different distributed environments (LAN vs WAN).
 - Can also maximize throughput
- Search algorithm
- How do we move inside the solution space?
 - Exhaustive search, heuristic algorithms (iterative improvement, simulated annealing, genetic,...)



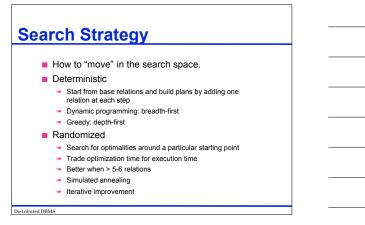


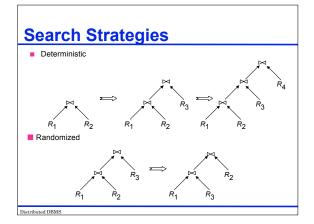


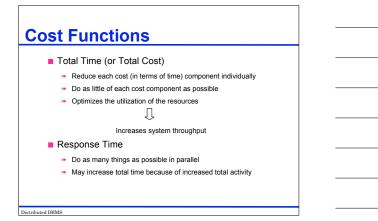




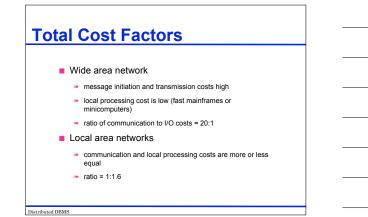








Total Cost Summation of all cost factors Total cost = CPU cost + I/O cost + communication cost = unit instruction cost * no.of instructions I/O cost = unit disk I/O cost * no. of disk I/Os communication cost = message initiation + transmission



Response Time

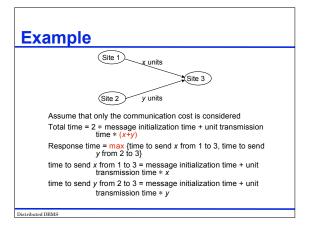
Elapsed time between the initiation and the completion of a query

Response time = CPU time + I/O time + communication time

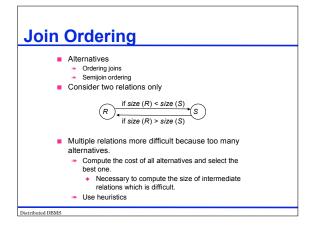
CPU time = unit instruction time * no. of sequential instructions

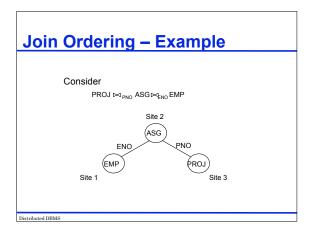
I/O time = unit I/O time * no. of sequential I/Os

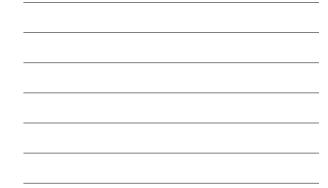
communication time = unit msg initiation time * no. of sequential msg + unit transmission time * no. of sequential bytes











Join Ordering – Example

Execution alternatives:

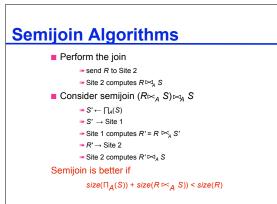
- 1. EMP \rightarrow Site 2 Site 2 computes EMP'=EMP $\mathsf{EMP'} \to \mathsf{Site}\; \mathsf{3}$ Site 3 computes EMP' by PROJ
- ASG → Site 1 Site 1 computes EMP'=EMP⊠ASG $\mathsf{EMP'} \to \mathsf{Site}\; \mathsf{3}$ Site 3 computes EMP'MPROJ
- 3. ASG \rightarrow Site 3 Site 3 computes ASG'=ASG ⋈ PROJ Site 2 computes PROJ'=PROJ⋈ASG ASG' → Site 1 Site 1 computes ASG ₩ EMP
- 4. PROJ → Site 2 PROJ' → Site 1 Site 1 computes PROJ'⊠ EMP
- 5. EMP \rightarrow Site 2 $\mathsf{PROJ} \to \mathsf{Site}\ \mathbf{2}$
 - Site 2 computes EMP 🖂 PROJ 🖂 ASG

Distributed DBMS

Semijoin Algorithms

- Consider the join of two relations: R[A] (located at site 1)
 - S[A] (located at site 2)
- Alternatives:
 - 1 Do the join $R \bowtie_A S$
 - 2 Perform one of the semijoin equivalents
 - $R \bowtie_A S \Leftrightarrow (R \bowtie_A S) \bowtie_A S$
 - $\Leftrightarrow \quad R \bowtie_A (S \bowtie_A R)$
 - $\Leftrightarrow \quad (R \Join_A S) \bowtie_A (S \bowtie_A R)$

Distributed DBMS



R* Algorithm

- Cost function includes local processing as well as transmission
- Considers only joins
- Exhaustive search
- Compilation

Distributed DBMS

 Published papers provide solutions to handling horizontal and vertical fragmentations but the implemented prototype does not

R* Algorithm Performing joins Ship whole larger data transfer smaller number of messages better if relations are small Fetch as needed number of messages = O(cardinality of external relation) data transfer per message is minimal better if relations are large and the selectivity is good

Distributed DBMS

R* Algorithm – **Vertical Partitioning & Joins**

1. Move outer relation tuples to the site of the inner relation

(a) Retrieve outer tuples

- (b) Send them to the inner relation site
- (c) Join them as they arrive
 - Total Cost = cost(retrieving qualified outer tuples)
 - + no. of outer tuples fetched * cost(retrieving qualified inner tuples)

 - + msg. cost * (no. outer tuples fetched * avg. outer tuple size) / msg. size

R* Algorithm – Vertical Partitioning & Joins

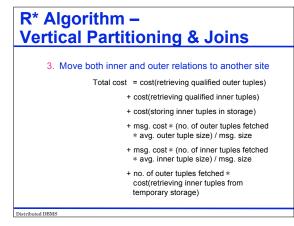
2. Move inner relation to the site of outer relation

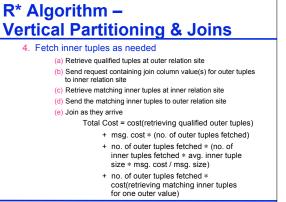
cannot join as they arrive; they need to be stored

Total Cost = cost(retrieving qualified outer tuples) + no. of outer tuples fetched *

cost(retrieving matching inner tuples from temporary storage)

- + cost(retrieving qualified inner tuples)+ cost(storing all qualified inner tuples
- in temporary storage)
- + msg. cost * (no. of inner tuples fetched * avg. inner tuple size) / msg. size
- Distributed DBMS

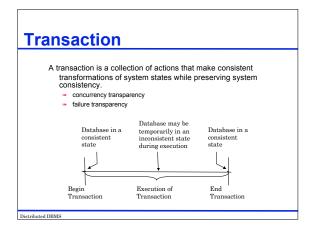






Outline

- Introduction
- Distributed DBMS Architecture
- Distributed Database Design
- Distributed Query Processing
- Distributed Concurrency Control
 - Transaction Concepts & Models
 - Serializability
 - Distributed Concurrency Control Protocols
- Distributed Reliability Protocols





Example Database

Consider an airline reservation example with the relations:

FLIGHT(<u>FNO, DATE</u>, SRC, DEST, STSOLD, CAP) CUST(<u>CNAME</u>, ADDR, BAL) FC(<u>FNO, DATE, CNAME</u>, SPECIAL)

Example Transaction

Distributed DBMS

Distributed DBMS

 Begin_transaction Reservation

 begin

 input(flight_no, date, customer_name);

 EXEC SQL
 UPDATE FLIGHT

 SET
 STSOLD = STSOLD + 1

 WHERE
 FNO = flight_no AND DATE = date;

 EXEC SQL
 INSERT

 INTO
 FC(FNO, DATE, CNAME, SPECIAL);

 VALUES
 (flight no, date, customer_name, null);

 output("reservation completed")

 Begin_transaction Reservation

 begin_ input(flight_no, date, customer_name); EXEC SQL

 SELECT
 STSOLD_CAP

 INTO
 temp1,temp2

 FROM
 FLIGHT

 WHERE
 FNO = flight_no AND DATE = date;

 if temp1 = temp2 then output(*no free seats*);

 Abort

 else

 EXEC SQL
 UPDATE

 EXEC SQL
 UPDATE

 SET
 STSOLD = STSOLD + 1 WHERE FNO = flight_no AND DATE = date;

 EXEC SQL
 INTO

 FC(FNO, DATE, CNAME, SPECIAL);

 VALUES (flight_no, date, customer_name, null);

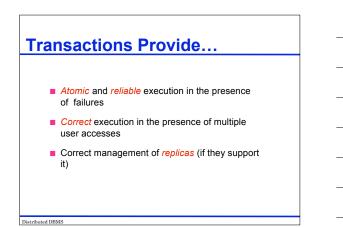
 Commit

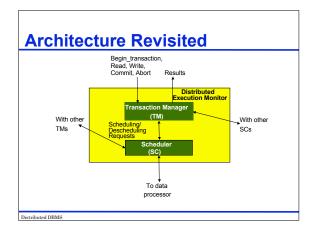
 output(*reservation completed*) endit

 end (_Reservation};

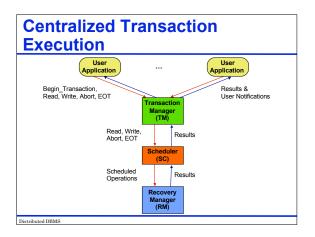
 Destributed DBMS

Properties of Transactions
Атомісіту
 all or nothing
 no violation of integrity constraints
SOLATION
 concurrent changes invisible È serializable
 committed updates persist
Distributed DBMS

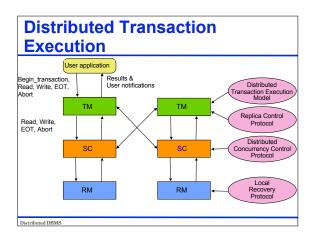














- The problem of synchronizing concurrent transactions such that the consistency of the database is maintained while, at the same time, maximum degree of concurrency is achieved.
- Anomalies:
 - Lost updates
 The effects of some transactions are not reflected on the database.
 - Inconsistent retrievals
 - A transaction, if it reads the same data item more than once, should always read the same value.

Serializable History

- Transactions execute concurrently, but the net effect of the resulting history upon the database is equivalent to some serial history.
- Equivalent with respect to what?
 - Conflict equivalence: the relative order of execution of the conflicting operations belonging to unaborted transactions in two histories are the same.
 - Conflicting operations: two incompatible operations (e.g., Read and Write) conflict if they both access the same data item.
 - Incompatible operations of each transaction is assumed
 - to conflict; do not change their execution orders.
 - If two operations from two different transactions conflict, the corresponding transactions are also said to conflict.

Distributed DBMS

Serializability in Distributed DBMS

- Somewhat more involved. Two histories have to be considered:
 - local histories
 - global history
- For global transactions (i.e., global history) to be serializable, two conditions are necessary:
 - Each local history should be serializable.
 - Two conflicting operations should be in the same relative order in all of the local histories where they appear together.

Distributed DBMS

Global Non-serializability

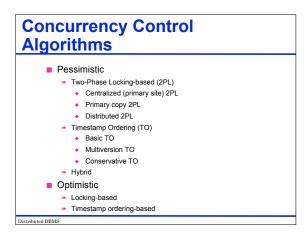
 $T_1: \begin{array}{c} \operatorname{Read}(x) \\ x \leftarrow x+5 \\ \operatorname{Write}(x) \\ \operatorname{Commit} \end{array}$

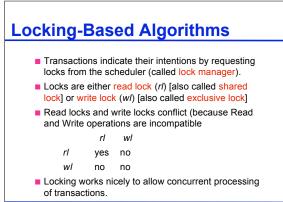
 $T_2: \operatorname{Read}(x) \\ x \leftarrow x*15 \\ \operatorname{Write}(x) \\ \widetilde{x}$

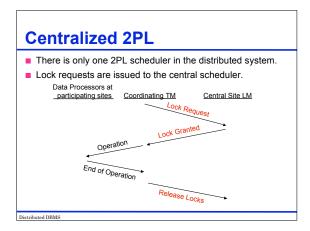
Write(x)Commit

The following two local histories are individually serializable (in fact serial), but the two transactions are not globally serializable.

$$\begin{split} LH_1 = & \{R_1(x), W_1(x), C_1, R_2(x), W_2(x), C_2\} \\ LH_2 = & \{R_2(x), W_2(x), C_2, R_1(x), W_1(x), C_1\} \end{split}$$



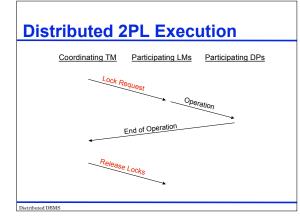




Distributed 2PL

Distributed DBMS

- 2PL schedulers are placed at each site. Each scheduler handles lock requests for data at that site.
- A transaction may read any of the replicated copies of item x, by obtaining a read lock on one of the copies of x. Writing into x requires obtaining write locks for all copies of x.



Timestamp Ordering

- **\mathbf{0}** Transaction (T_i) is assigned a globally unique timestamp $ts(T_i)$.
- O Transaction manager attaches the timestamp to all
- operations issued by the transaction.
- Seach data item is assigned a write timestamp (wts) and a read timestamp (rts): rts(x) = largest timestamp of any read on x
 - wts(x) = largest timestamp of any read on x
- **G** Conflicting operations are resolved by timestamp order. Basic T/O: for $R_j(x)$ if $ts(T_j) < wts(x)$ then reject $R_j(x)$ else accept $R_j(x)$ for $W_i(x)$ if $ts(T_i) < rts(x)$ and $ts(T_i) < wts(x)$ then reject $W_i(x)$ else accept $W_i(x)$

 $wts(x) \leftarrow ts(T_i)$

Distributed DBMS

 $rts(x) \leftarrow ts(T_j)$

Outline

Distributed DBMS

- Introduction
- Distributed DBMS Architecture
- Distributed Database Design
- Distributed Query Processing
- Distributed Concurrency Control
- Distributed Reliability Protocols
 - Distributed Commit ProtocolsDistributed Recovery Protocols

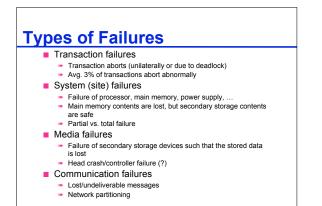
 Problem:

 How to maintain

 atomicity

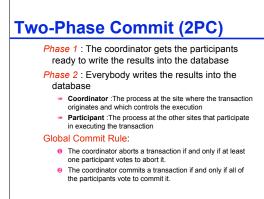
 durability

 properties of transactions

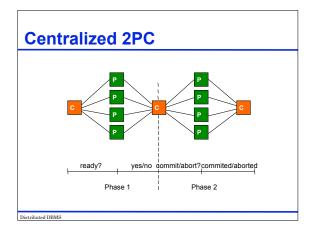


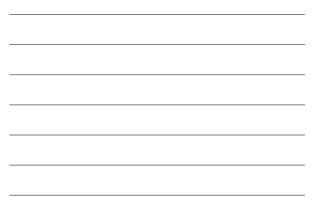


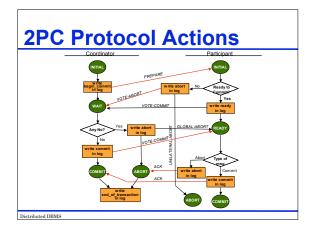
- Commit protocols
 - How to execute commit command for distributed transactions.
 Issue: how to ensure atomicity and durability?
- Termination protocols
 - If a failure occurs, how can the remaining operational sites deal with it. Non-blocking : the occurrence of failures should not force the
 - sites to wait until the failure is repaired to terminate the transaction.
- Recovery protocols
 - When a failure occurs, how do the sites where the failure occurred deal with it.
 - Independent : a failed site can determine the outcome of a transaction without having to obtain remote information.
- Independent recovery ⇒ non-blocking termination













Problem With 2PC

- Blocking
 - Ready implies that the participant waits for the coordinator
 - If coordinator fails, site is blocked until recovery
 - Blocking reduces availability
- Independent recovery is not possible
- However, it is known that:
 - Independent recovery protocols exist only for single site failures; no independent recovery protocol exists which is resilient to multiple-site failures.
- So we search for these protocols 3PC