Distributed Database Management Systems

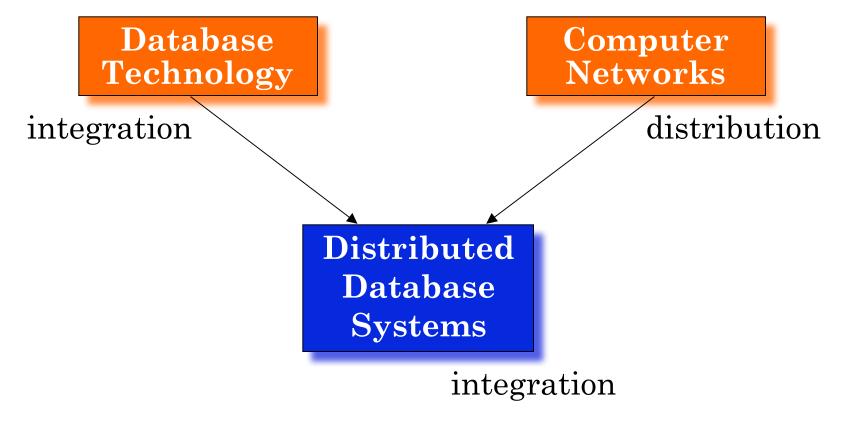


- Introduction
- Distributed DBMS Architecture
- Distributed Database Design
- Distributed Query Processing
- Distributed Concurrency Control
- Distributed Reliability Protocols

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

Motivation



 $integration \neq centralization$

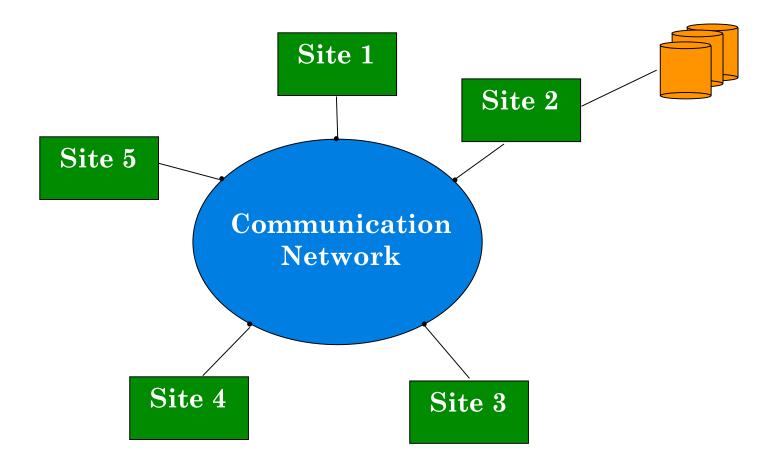
What is a Distributed Database System?

A distributed database (DDB) is a collection of multiple, *logically interrelated* databases distributed over a *computer network*.

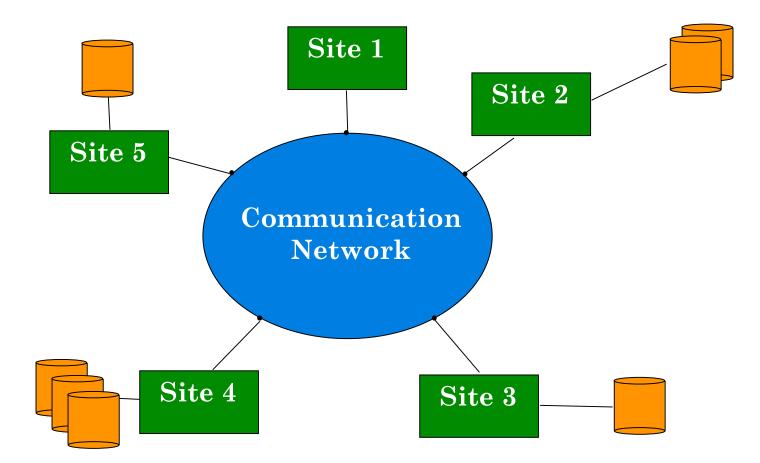
A distributed database management system (D–DBMS) is the software that manages the DDB and provides an access mechanism that makes this distribution transparent to the users.

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

Centralized DBMS on Network



Distributed DBMS Environment



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 Promises

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

Transparency

- Transparency is the separation of the higher level semantics of a system from the lower level implementation issues.
- Fundamental issue is to provide

data independence

- in the distributed environment
 - Network (distribution) transparency
 - Replication transparency
 - Fragmentation transparency
 - horizontal fragmentation: selection
 - vertical fragmentation: projection
 - hybrid

Example

EMP			ASG			
ENO	ENAME	TITLE	ENO	PNO	RESP	DUR
E1 E2 E3 E4 E5 E6 E7 E8	J. Doe M. Smith A. Lee J. Miller B. Casey L. Chu R. Davis J. Jones	Elect. Eng. Syst. Anal. Mech. Eng. Programmer Syst. Anal. Elect. Eng. Mech. Eng. Syst. Anal.	E1 E2 E3 E3 E3 E4 E5 E6 E7 E7 E8	P1 P1 P2 P3 P4 P2 P2 P4 P3 P5 P3	Manager Analyst Analyst Consultant Engineer Programmer Manager Manager Engineer Engineer Manager	$ \begin{array}{r} 12 \\ 24 \\ 6 \\ 10 \\ 48 \\ 18 \\ 24 \\ 48 \\ 36 \\ 23 \\ 40 \\ \end{array} $

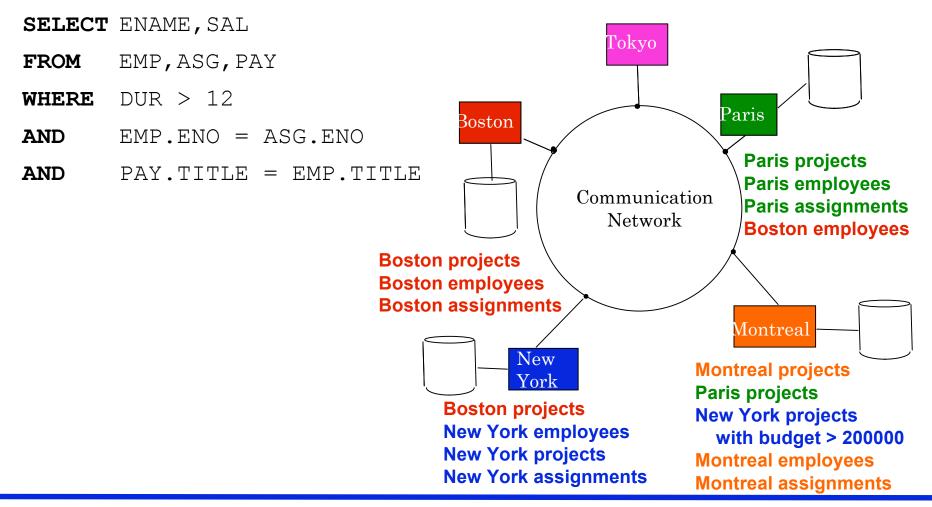
PROJ

PNO	PNAME	BUDGET
P3	Instrumentation Database Develop. CAD/CAM Maintenance	$\begin{array}{c} 150000\\ 135000\\ 250000\\ 310000 \end{array}$

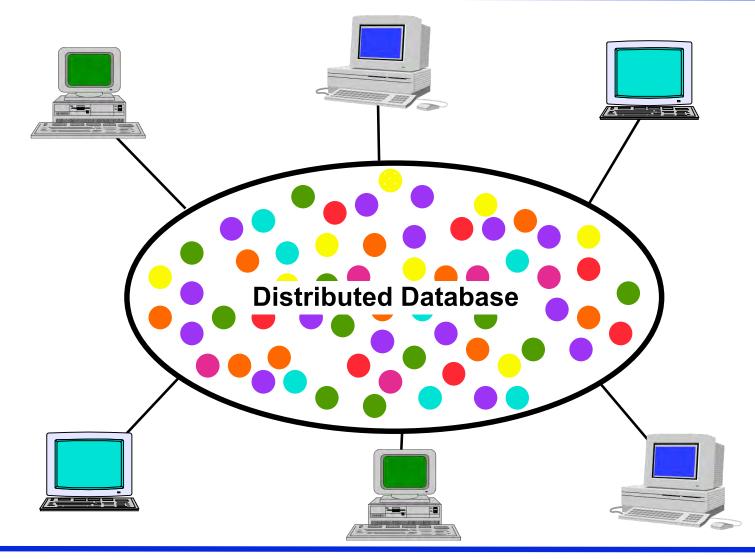
PAY

TITLE	SAL
Elect. Eng. Syst. Anal. Mech. Eng. Programmer	$\begin{array}{c} 40000\\ 34000\\ 27000\\ 24000 \end{array}$

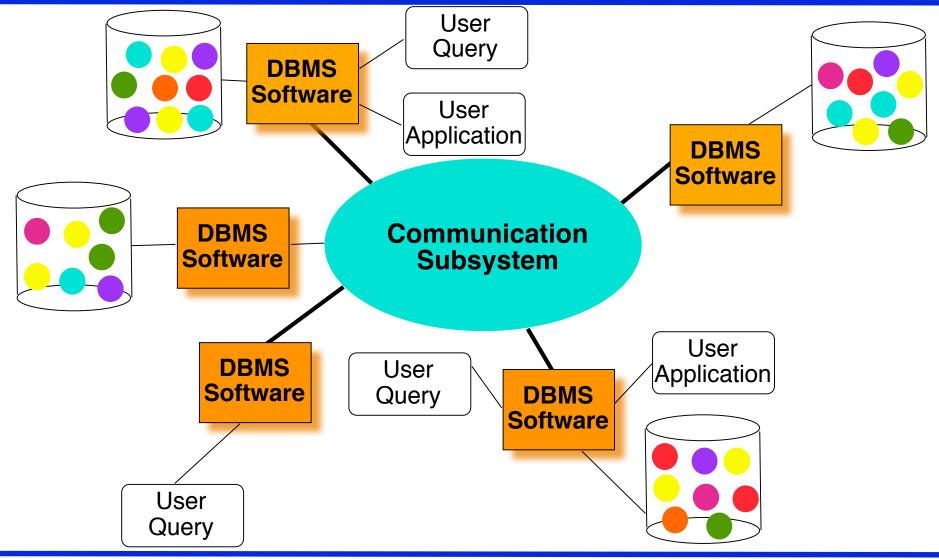
Transparent Access



Distributed Database – User View



Distributed DBMS - Reality



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

Parallelism Requirements

- Have as much of the data required by each application at the site where the application executes
 - Full replication
- How about updates?
 - Updates to replicated data requires implementation of distributed concurrency control and commit protocols

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 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

Distributed DBMS Issues

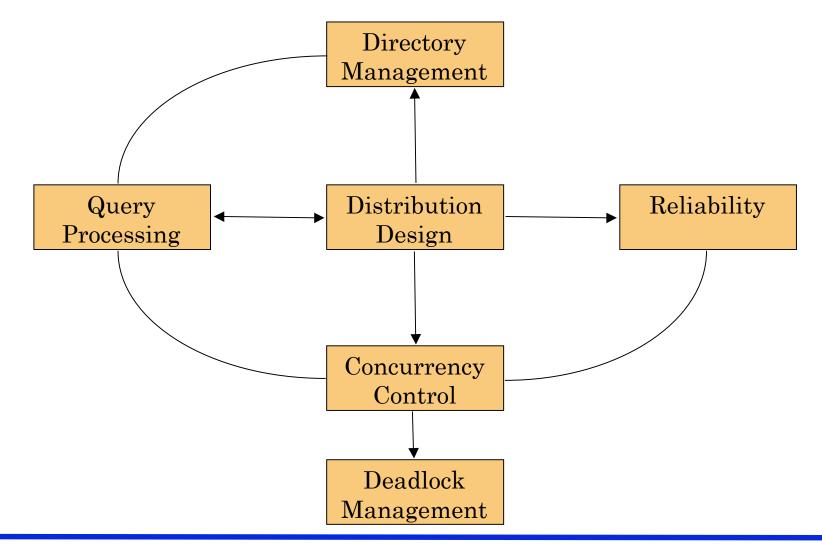
Concurrency Control

- synchronization of concurrent accesses
- consistency and isolation of transactions' effects
- deadlock management

Reliability

- how to make the system resilient to failures
- atomicity and durability

Relationship Between Issues



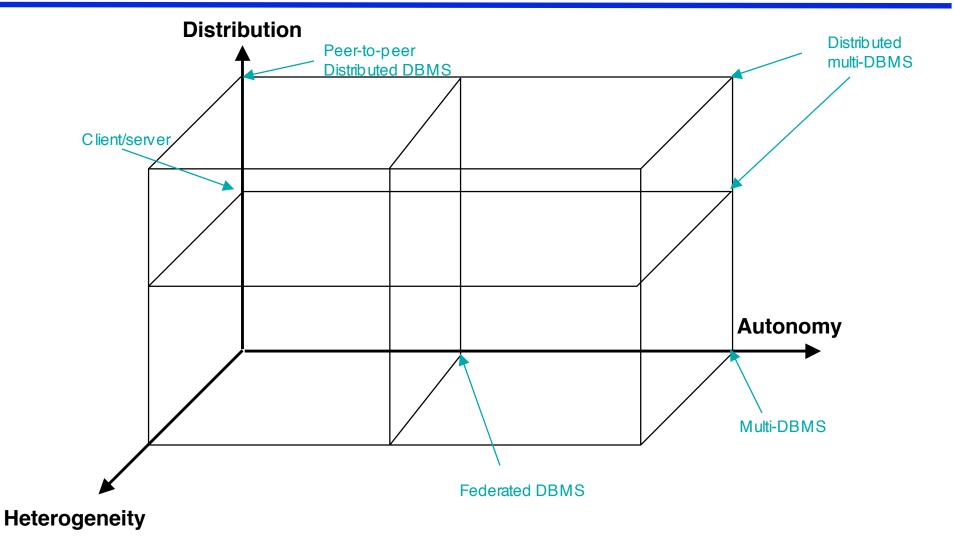
Outline

Introduction

Distributed DBMS Architecture

- Implementation Alternatives
- Component Architecture
- Distributed Database Design
- Distributed Query Processing
- Distributed Concurrency Control
- Distributed Reliability Protocols

DBMS Implementation Alternatives



Dimensions of the Problem

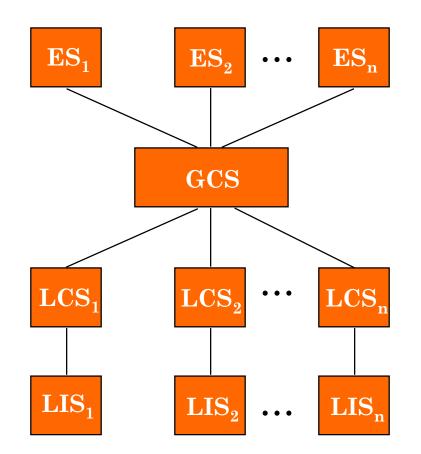
Distribution

- 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

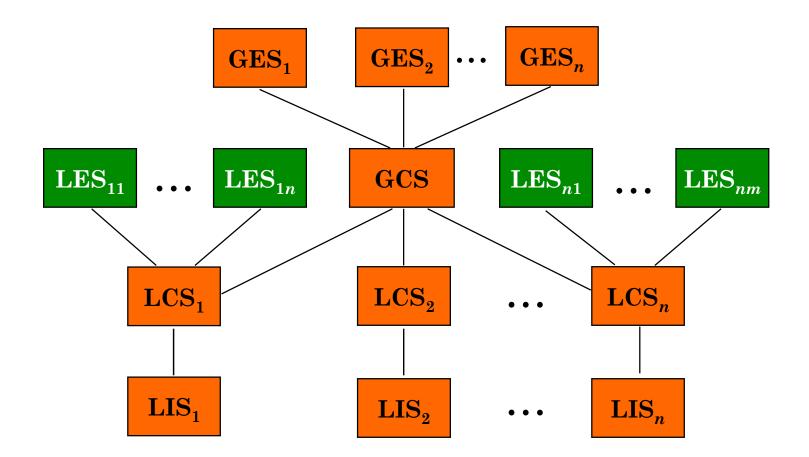
Autonomy

- Not well understood and most troublesome
- Various versions
 - Design autonomy: Ability of a component DBMS to decide on 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.

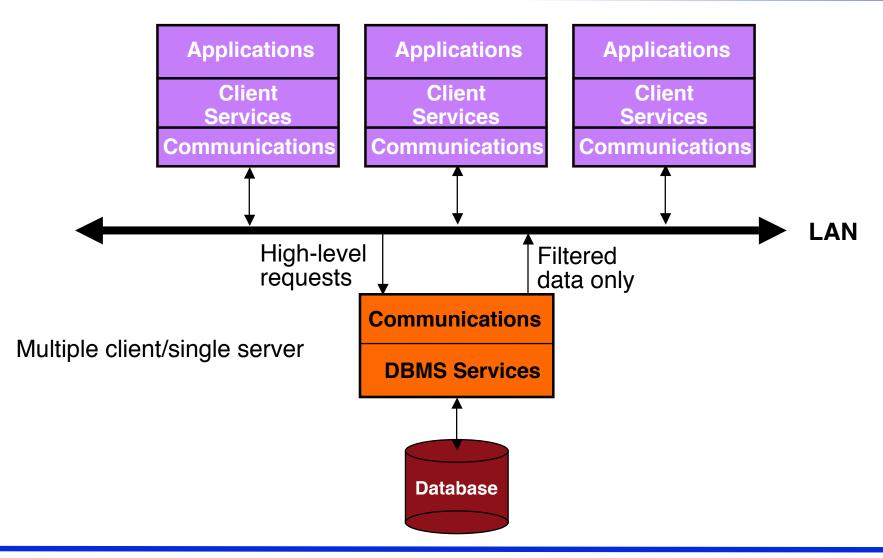
Datalogical Distributed DBMS Architecture



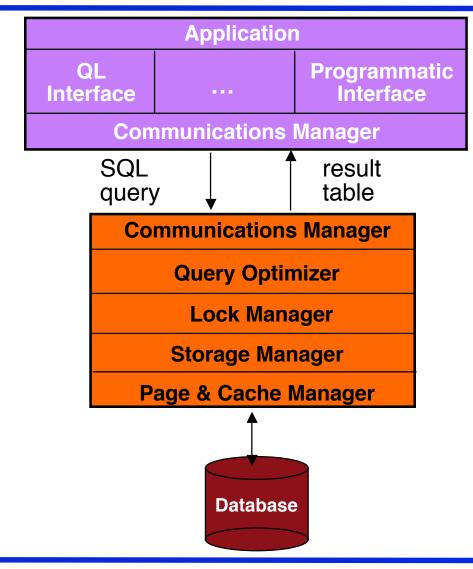
Datalogical Multi-DBMS Architecture



Clients/Server



Task Distribution



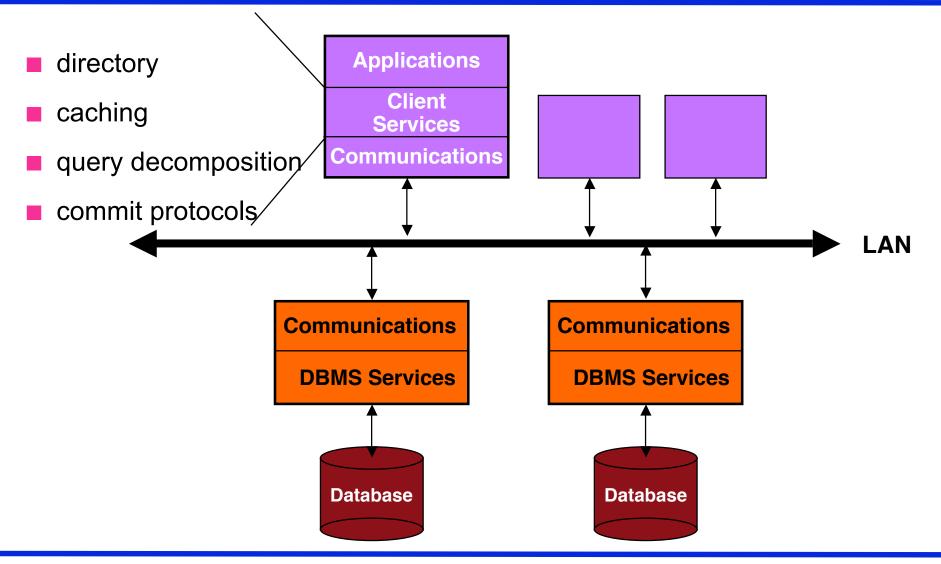
Advantages of Client-Server Architectures

- 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

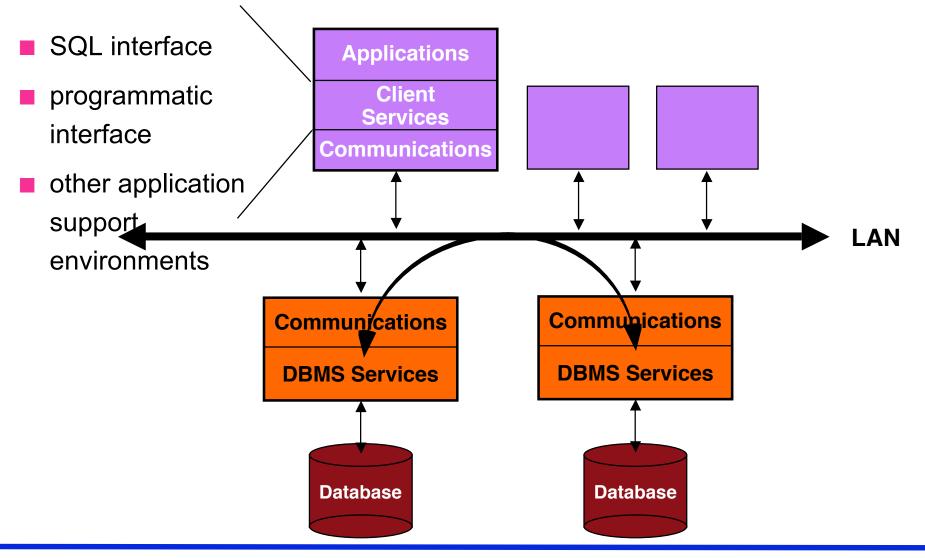
Problems With Multiple-Client/Single Server

- Server forms bottleneck
- Server forms single point of failure
- Database scaling difficult

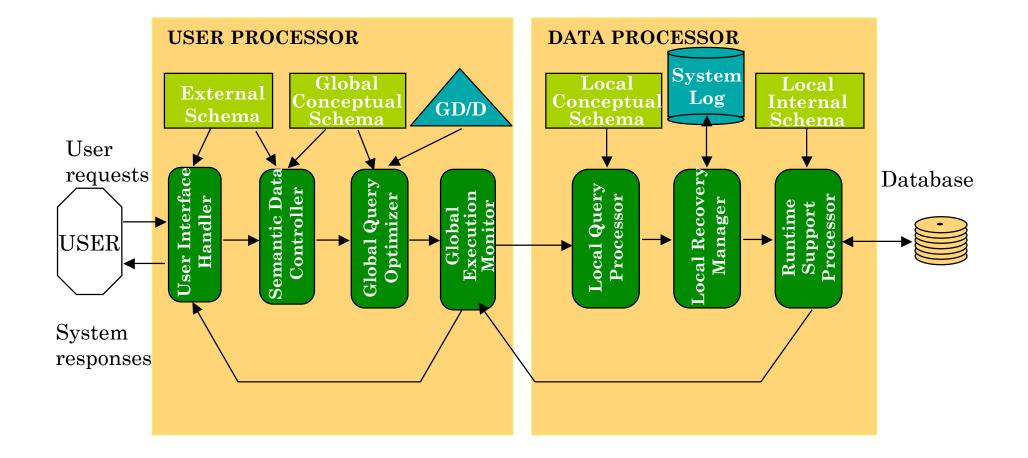
Multiple Clients/Multiple Servers



Server-to-Server



Peer-to-Peer Component Architecture





- Introduction
- Distributed DBMS Architecture
- Distributed Database Design
 - **Fragmentation**
 - Data Placement
- Distributed Query Processing
- Distributed Concurrency Control
- Distributed Reliability Protocols

Design Problem

- In the general setting :
 - Making decisions about the placement of data and programs across the sites of a computer network as well as possibly designing the network itself.
- In Distributed DBMS, the placement of applications entails
 - placement of the distributed DBMS software; and
 - placement of the applications that run on the database

Distribution Design

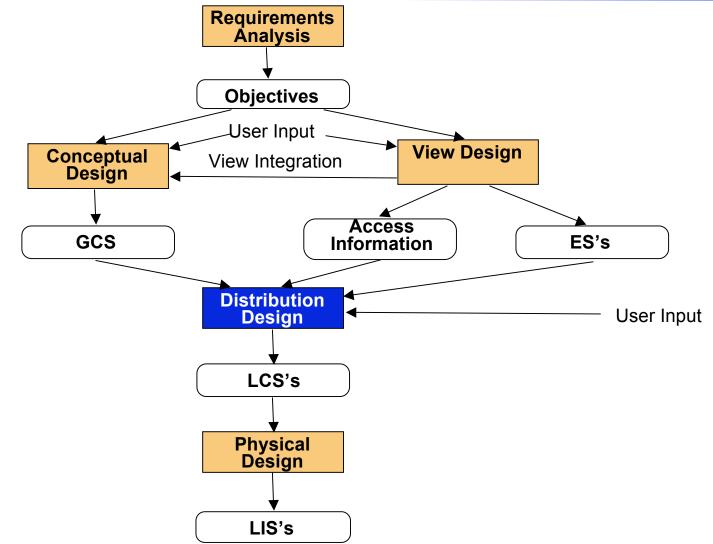
Top-down

- mostly in designing systems from scratch
- mostly in homogeneous systems

Bottom-up

when the databases already exist at a number of sites

Top-Down Design



Distribution Design

Fragmentation

- Localize access
- Horizontal fragmentation
- Vertical fragmentation
- Hybrid fragmentation
- Distribution
 - Placement of fragments on nodes of a network

Horizontal Fragmentation

- PROJ₁ : projects with budgets less than \$200,000
- PROJ₂ : projects with budgets greater than or equal to \$200,000

PNO	PNAME	BUDGET	LOC
	Instrumentation Database Develop. CAD/CAM Maintenance CAD/CAM	135000	

PROJ₁

PNO	PNAME	BUDGET	LOC
P1	Instrumentation	150000 l	Montreal
P2	Database Develop.	135000 I	New York

PROJ₂

PROJ

PNO	PNAME	BUDGET	LOC
P3	CAD/CAM	250000	New York
P4	Maintenance	310000	Paris
P5	CAD/CAM	500000	Boston

Vertical Fragmentation

PROJ ₁ :	information about
	project budgets

PROJ₂: information about project names and locations

PNO	PNAME I	BUDGET	LOC
	Instrumentation Database Develop. CAD/CAM Maintenance CAD/CAM	135000	

l	PROJ		
	PNO	BUDGET	
	P1 P2 P3 P4 P5	150000 135000 250000 310000 500000	

PROJ₂

PROJ

PNO	PNAME	LOC
P2	Instrumentation Database Develop. CAD/CAM Maintenance CAD/CAM	Montreal New York New York Paris Boston

Correctness of Fragmentation

Completeness

- Decomposition of relation *R* into fragments $R_1, R_2, ..., R_n$ is complete iff each data item in *R* can also be found in some R_i
- Reconstruction
 - If relation *R* is decomposed into fragments $R_1, R_2, ..., R_n$, then there should exist some relational operator ∇ such that

$$R = \nabla_{1 \le i \le n} R_i$$

Disjointness

If relation *R* is decomposed into fragments $R_1, R_2, ..., R_n$, and data item d_i is in R_j , then d_i should not be in any other fragment R_k ($k \neq j$).

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 $\frac{read - only queries}{update queries} \ge 1$ replication is advantageous, otherwise replication may cause problems

Fragment 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)

Allocation Model

General Form

min(Total Cost)

subject to

response time constraint storage constraint processing constraint

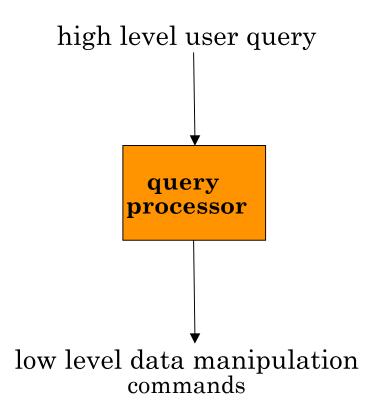
Decision Variable

$$x_{ij} = \begin{cases} 1 \text{ if fragment } F_i \text{ is stored at site } S_j \\ 0 \text{ otherwise} \end{cases}$$

Outline

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

Query Processing



Distributed DBMS

Query Processing Components

- Query language that is used
 - SQL: "intergalactic dataspeak"
- Query execution methodology
 - The steps that one goes through in executing high-level (declarative) user queries.
- Query optimization
 - How do we determine the "best" execution plan?

Selecting Alternatives

SELECT	ENAME
FROM	EMP,ASG
WHERE	EMP.ENO = ASG.ENO
AND	DUR > 37

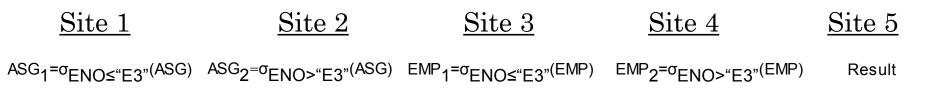
Strategy 1

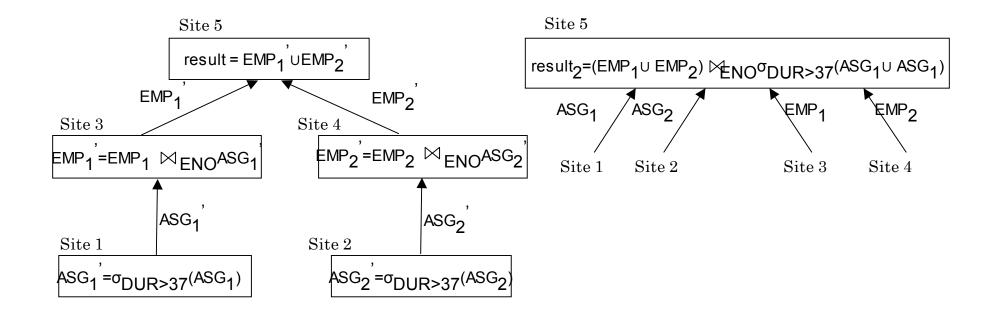
 $\Pi_{\text{ENAME}}(\sigma_{\text{DUR>37} \land \text{EMP}.\text{ENO=ASG}.\text{ENO}} (\text{EMP} \times \text{ASG}))$ Strategy 2

 $\Pi_{\text{ENAME}}(\text{EMP} \bowtie_{\text{ENO}} (\sigma_{\text{DUR>37}}(\text{ASG})))$

Strategy 2 avoids Cartesian product, so is "better"

What is the Problem?





Cost of Alternatives

Assume:

- *size*(EMP) = 400, *size*(ASG) = 1000
- tuple access cost = 1 unit; tuple transfer cost = 10 units

Strategy 1

0	produce ASG': (10+10)*tuple access cost	20	
2	transfer ASG' to the sites of EMP: (10+10)*tuple transfer cost	200	
8	produce EMP': (10+10) *tuple access cost*2	40	
4	transfer EMP' to result site: (10+10) *tuple transfer cost	200	
	Total cost	460	
Strategy 2			
0	transfer EMP to site 5:400*tuple transfer cost	4,000	
2	transfer ASG to site 5 :1000*tuple transfer cost	10,000	
8	produce ASG':1000*tuple access cost	1,000	
4	join 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

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

- \rightarrow compilation \Rightarrow 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

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

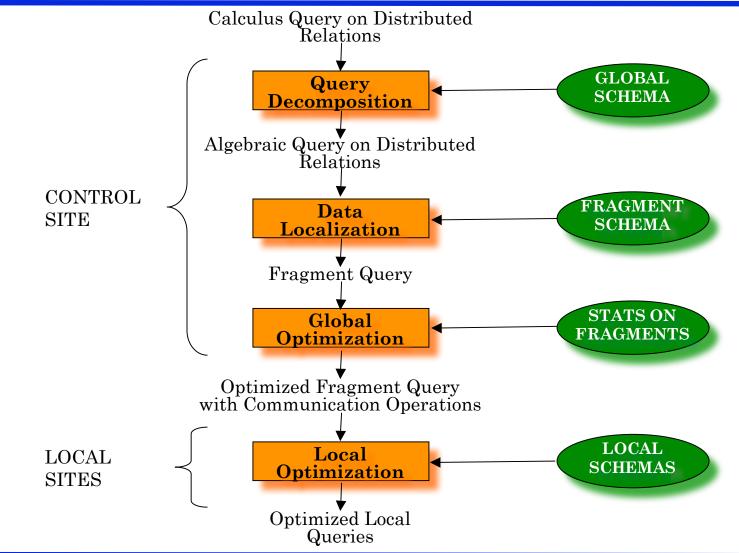
Centralized

- 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

Query Optimization Issues – Network Topology

- Wide area networks (WAN) point-to-point
 - characteristics
 - low bandwidth
 - low speed
 - high protocol overhead
 - communication cost will dominate; ignore all other cost factors
 - global schedule to minimize communication cost
 - Iocal 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

Distributed Query Processing Methodology



Distributed DBMS

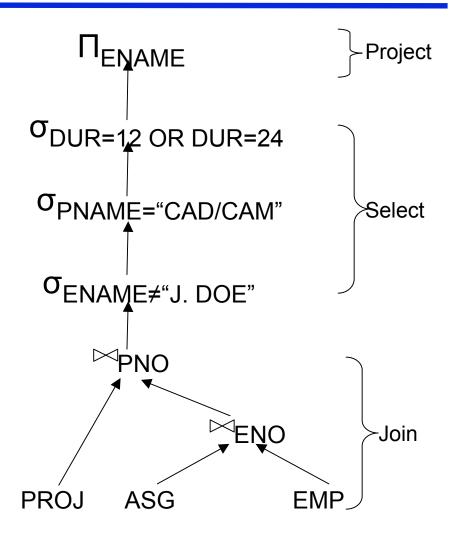
Step 1 – Query Decomposition

Input: Calculus query on global relations

- Normalization
 - manipulate query quantifiers and qualification
- Analysis
 - detect and reject "incorrect" queries
 - possible for only a subset of relational calculus
- Simplification
 - eliminate redundant predicates
- Restructuring
 - → calculus query \Rightarrow algebraic query
 - more than one translation is possible
 - use transformation rules

Restructuring

- Convert relational calculus to relational algebra Make use of query trees Example Find the names of employees other than J. Doe who worked on the CAD/CAM project for either 1 or 2 years. SELECT ENAME FROM EMP, ASG, PROJ EMP.ENO = ASG.ENOWHERE ASG.PNO = PROJ.PNOAND AND ENAME ≠ "J. Doe" PNAME = "CAD/CAM"AND
 - **AND** (DUR = 12 **OR** DUR = 24)



Restructuring – Transformation Rules (Examples)

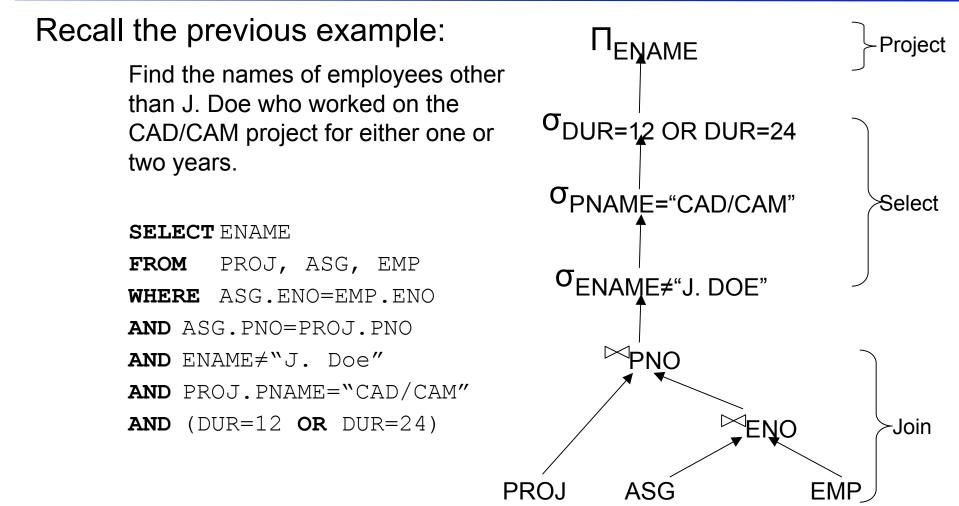
- Commutativity of binary operations
 - $\implies R \times S \Leftrightarrow S \times R$
 - $\implies R \bowtie S \Leftrightarrow S \bowtie R$
 - $\implies R \cup S \Leftrightarrow S \cup R$
- Associativity of binary operations

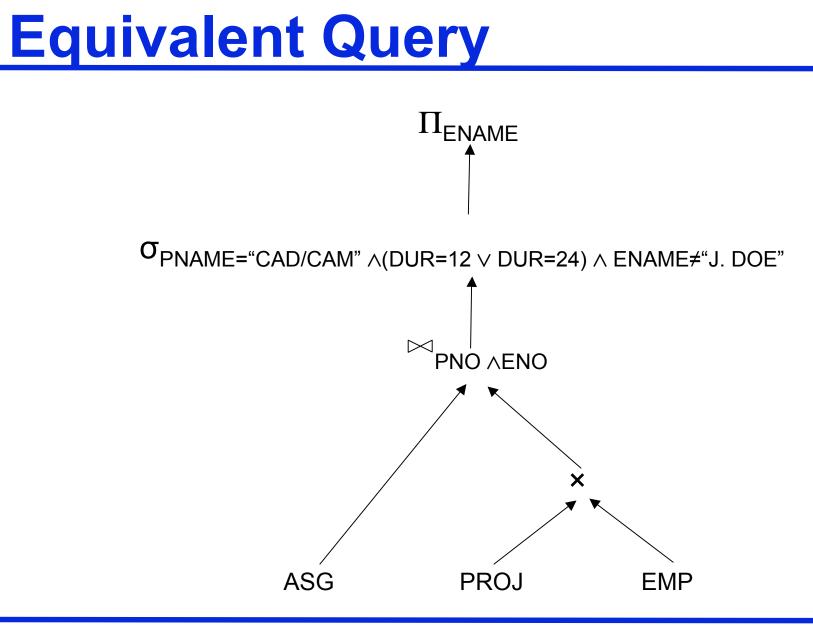
$$(R \times S) \times T \Leftrightarrow R \times (S \times T)$$

- $\implies (R \bowtie S) \bowtie T \Leftrightarrow R \bowtie (S \bowtie T)$
- Idempotence of unary operations

- [™] $\sigma_{p_1(A_1)}(\sigma_{p_2(A_2)}(R)) = \sigma_{p_1(A_1)} \land p_2(A_2)(R)$ where *R*[*A*] and *A*' ⊆ *A*, *A*" ⊆ *A* and *A*' ⊆ *A*"
- Commuting selection with projection

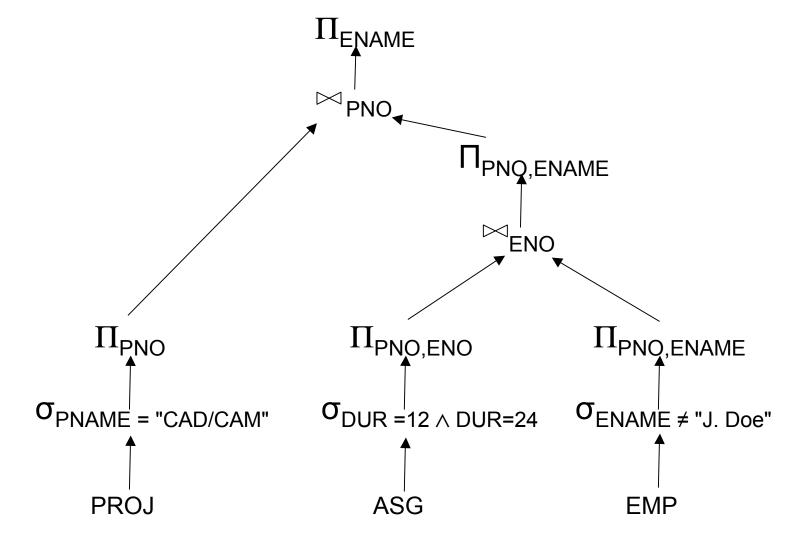
Example





Distributed DBMS

Restructuring



Distributed DBMS

Step 2 – Data Localization

Input: Algebraic query on distributed relations

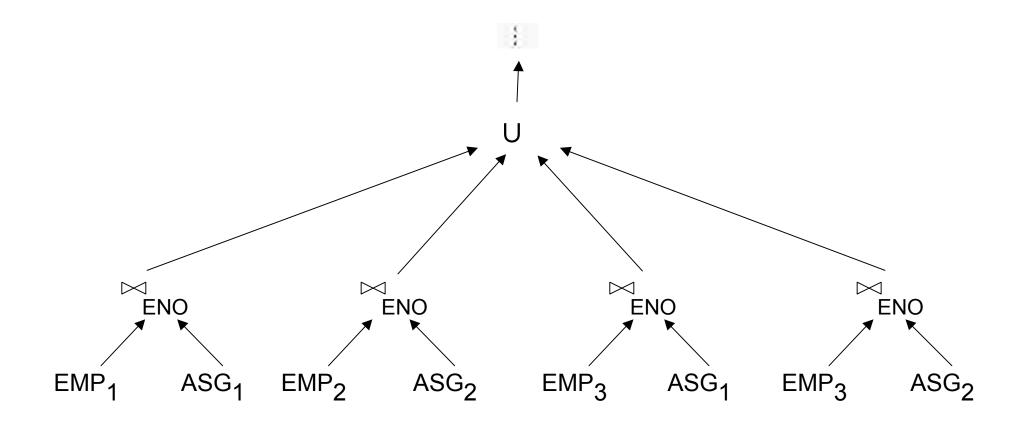
- Determine which fragments are involved
- Localization program
 - substitute for each global query its materialization program
 - optimize

Example

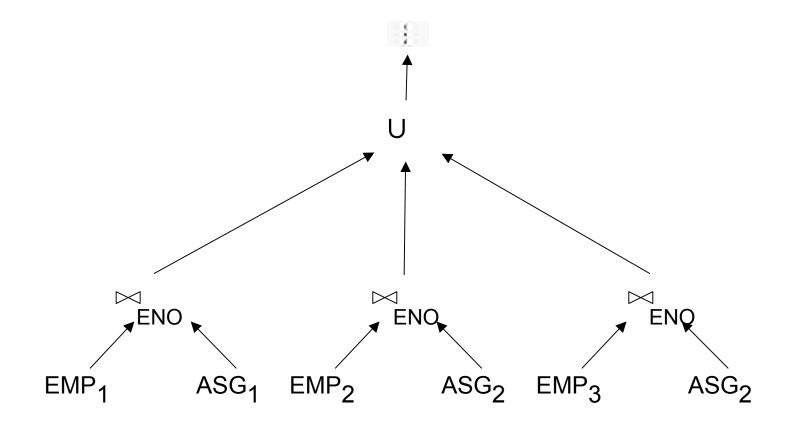
II_{ENAME} Assume EMP is fragmented into EMP_1 , EMP_2 , ODUR=12 OR DUR=24 EMP_3 as follows: • $EMP_1 = \sigma_{ENO \leq "E3"}(EMP)$ σ_{PNAME="CAD/CAM"} • EMP₂= $\sigma_{\text{"E3"} < \text{ENO} \leq \text{"E6"}}$ (EMP) • EMP₃= $\sigma_{FNO \geq "F6"}(EMP)$ σ_{ENAME≠"J. DOE"} ➡ ASG fragmented into ASG₁ and ASG₂ as follows: \sim • $ASG_1 = \sigma_{FNO \leq "F3"}(ASG)$ **PNO** • ASG₂= $\sigma_{FNO>"F3"}(ASG)$ \bowtie ENO Replace EMP by (EMP₁ \cup EMP₂ \cup EMP₃) PROJ and ASG by $(ASG_1 \cup ASG_2)$ in any query EMP1 EMP2 EMP3 ASG1 ASG₂

Distributed DBMS

Provides Parallellism



Eliminates Unnecessary Work



Step 3 – Global Query Optimization

Input: Fragment query

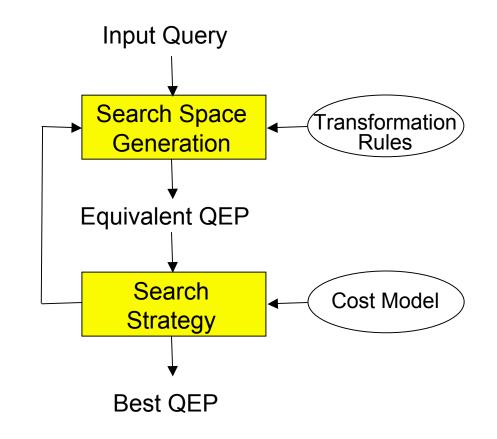
- Find the *best* (not necessarily optimal) global schedule
 - Minimize a cost function
 - Distributed join processing
 - Bushy vs. linear trees
 - Which relation to ship where?
 - Ship-whole vs ship-as-needed
 - Decide on the use of semijoins
 - Semijoin saves on communication at the expense of more 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,...)

Query Optimization Process

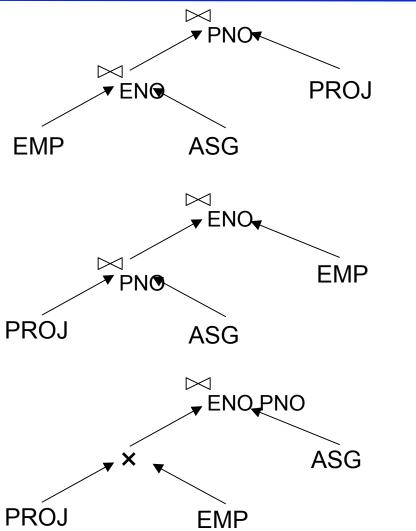


Search Space

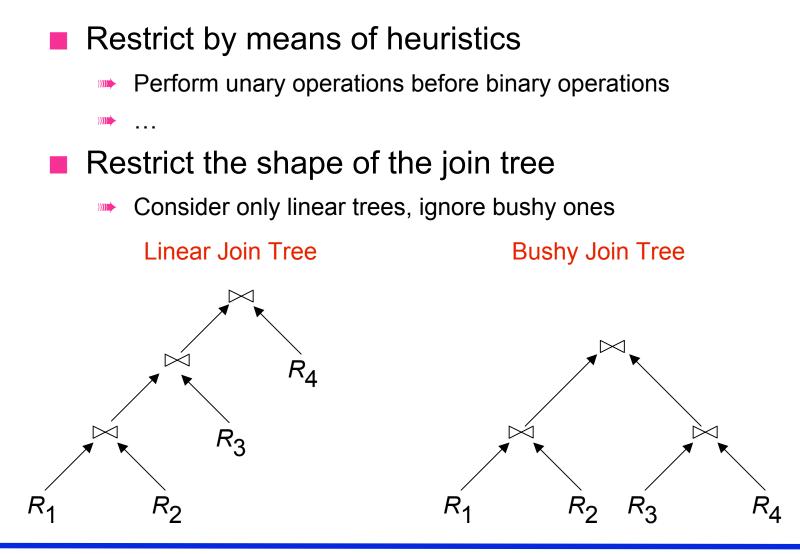
- Search space characterized by alternative execution plans
- Focus on join trees
- For N relations, there are O(N!) equivalent join trees that can be obtained by applying commutativity and associativity rules



- FROM EMP, ASG, PROJ
- WHERE EMP.ENO=ASG.ENO
- AND ASG.PNO=PROJ.PNO



Search Space



Distributed DBMS

Search Strategy

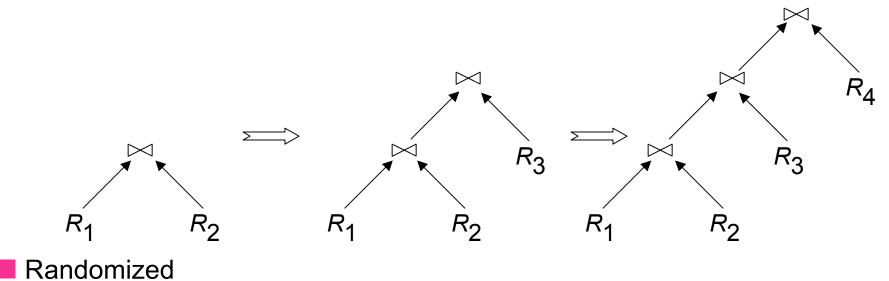
How to "move" in the search space.

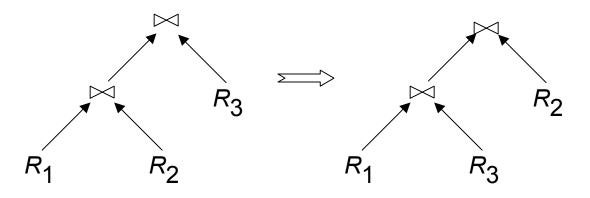
Deterministic

- Start from base relations and build plans by adding one relation at each step
- Dynamic programming: breadth-first
- Greedy: depth-first
- Randomized
 - Search for optimalities around a particular starting point
 - Trade optimization time for execution time
 - Better when > 5-6 relations
 - Simulated annealing
 - Iterative improvement

Search Strategies

Deterministic





Cost Functions

Total Time (or Total Cost)

- Reduce each cost (in terms of time) component individually
- Do as little of each cost component as possible
- Optimizes the utilization of the resources



- Response Time
 - Do as many things as possible in parallel
 - May increase total time because of increased total activity



Summation of all cost factors

Total cost	= CPU cost + I/O cost + communication
cost	

- CPU 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

Total Cost Factors

- Wide area network
 - message initiation and transmission costs high
 - Iocal processing cost is low (fast mainframes or minicomputers)
 - ratio of communication to I/O costs = 20:1
- Local area networks
 - communication and local processing costs are more or less equal
 - ratio = 1:1.6

Response Time

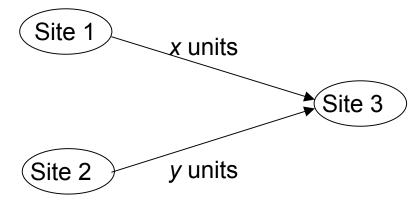
Elapsed time between the initiation and the completion of a query

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





Assume that only the communication cost is considered

- Total time = 2 * message initialization time + unit transmission time * (x+y)
- Response time = max {time to send x from 1 to 3, time to send y from 2 to 3}
- time to send x from 1 to 3 = message initialization time + unit transmission time * x

time to send *y* from 2 to 3 = message initialization time + unit transmission time * *y*

Join Ordering

Alternatives

- Ordering joins
- Semijoin ordering
- Consider two relations only

$$R \xrightarrow{\text{if size } (R) < \text{size } (S)} S$$

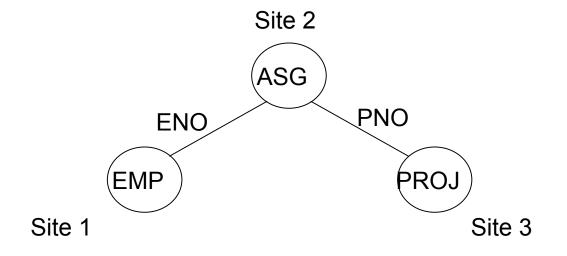
$$if size (R) > size (S)$$

- Multiple relations more difficult because too many alternatives.
 - Compute the cost of all alternatives and select the best one.
 - Necessary to compute the size of intermediate relations which is difficult.
 - Use heuristics

Join Ordering – Example

Consider

PROJ ⋈ _{PNO} ASG ⋈_{ENO} EMP



Join Ordering – Example

Execution alternatives:

- 1. EMP \rightarrow Site 2 Site 2 computes EMP'=EMP $\triangleright ASG$ EMP' \rightarrow Site 3 Site 3 computes EMP' $\triangleright PROJ$
- 3. ASG \rightarrow Site 3 Site 3 computes ASG'=ASG \bowtie PROJ ASG' \rightarrow Site 1 Site 1 computes ASG' \bowtie EMP
- 5. EMP \rightarrow Site 2 PROJ \rightarrow Site 2 Site 2 computes EMP \bowtie PROJ \bowtie ASG

- 2. ASG \rightarrow Site 1 Site 1 computes EMP'=EMP \bowtie ASG EMP' \rightarrow Site 3 Site 3 computes EMP' \bowtie PROJ
- 4. PROJ → Site 2
 Site 2 computes PROJ'=PROJ ⋈ ASG
 PROJ' → Site 1
 Site 1 computes PROJ' ⋈ EMP

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
$$R \bowtie_A S$$

$$R \bowtie_{A} S \Leftrightarrow (R \bowtie_{A} S) \bowtie_{A} S$$
$$\Leftrightarrow R \bowtie_{A} (S \bowtie_{A} R)$$
$$\Leftrightarrow (R \bowtie_{A} S) \bowtie_{A} (S \bowtie_{A} R)$$

Semijoin Algorithms

Perform the join

- send R to Site 2
- \blacksquare Site 2 computes $R \bowtie_A S$
- Consider semijoin $(R \bowtie_A S) \bowtie_A S$
 - $\Longrightarrow S' \leftarrow \prod_A(S)$
 - \Longrightarrow S' \rightarrow Site 1
 - ⇒ Site 1 computes $R' = R \bowtie_A S'$
 - $\gg R' \rightarrow Site 2$
 - \blacksquare Site 2 computes $R' \bowtie_A S$

Semijoin is better if

 $size(\Pi_A(S)) + size(R \bowtie_A S)) < size(R)$

R* Algorithm

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

- 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

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

3. Move both inner and outer relations to another site

Total cost = cost(retrieving qualified outer tuples)

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

4. Fetch inner tuples as needed

- (a) Retrieve qualified tuples at outer relation site
- (b) Send request containing join column value(s) for outer tuples to inner relation site
- (c) Retrieve matching inner tuples at inner relation site
- (d) Send the matching inner tuples to outer relation site
- (e) Join as they arrive

Total Cost = cost(retrieving qualified outer tuples)

- + msg. cost * (no. of outer tuples fetched)
- + no. of outer tuples fetched * (no. of inner tuples fetched * avg. inner tuple size * msg. cost / msg. size)
- + no. of outer tuples fetched * cost(retrieving matching inner tuples for one outer value)

Step 4 – Local Optimization

Input: Best global execution schedule

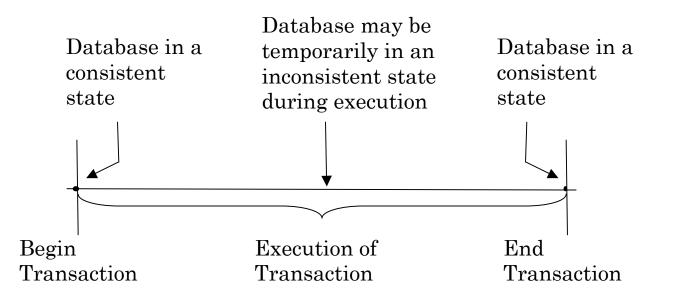
- Select the best access path
- Use the centralized optimization techniques

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

Transaction

- A transaction is a collection of actions that make consistent transformations of system states while preserving system consistency.
 - concurrency transparency
 - failure transparency



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

```
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")

end . {Reservation}
```

Termination of Transactions

```
Begin_transaction Reservation
begin
input(flight_no, date, customer_name);
                           STSOLD,CAP
FXFC SQL
               SFI FCT
               INTO
                           temp1,temp2
               FROM
                           FI IGHT
               WHERE
                           FNO = flight no AND DATE = date;
if temp1 = temp2 then
output("no free seats");
Abort
else
FXFC 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);
  Commit
  output("reservation completed")
  endif
end . {Reservation}
```

Properties of Transactions



all or nothing



no violation of integrity constraints

SOLATION

concurrent changes invisible È serializable

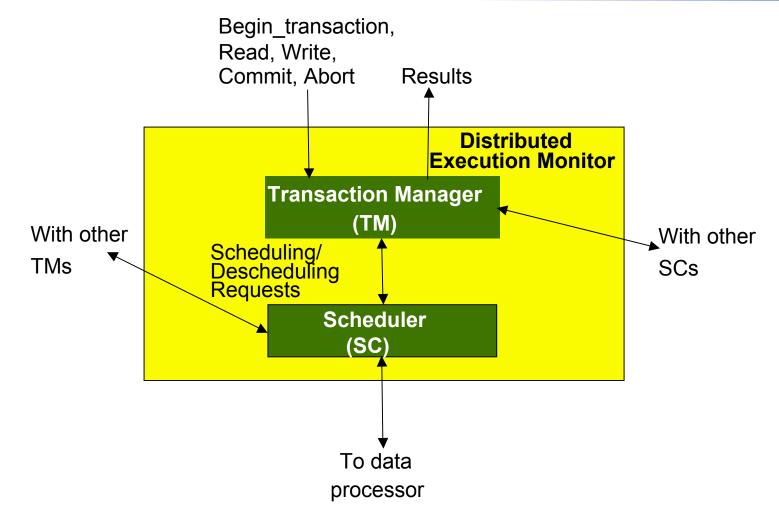
DURABILITY

committed updates persist

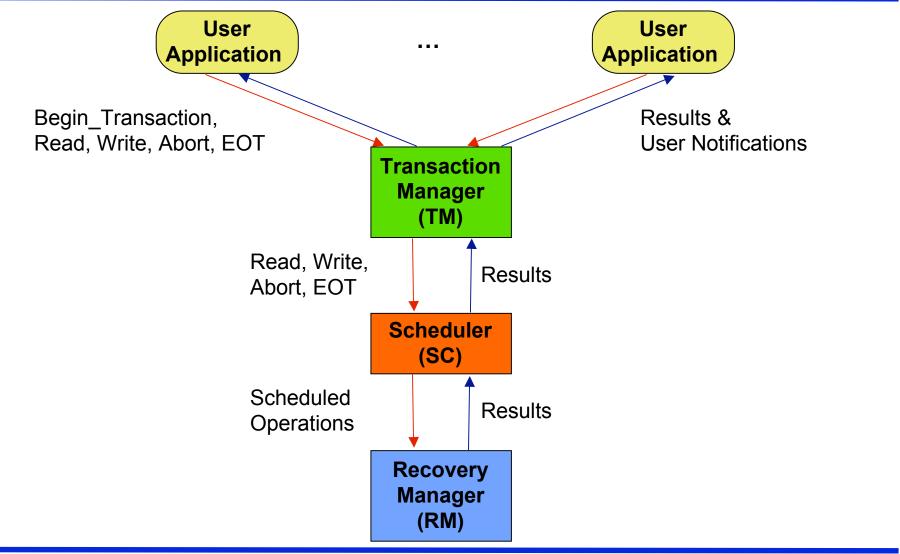
Transactions Provide...

- Atomic and reliable execution in the presence of failures
- Correct execution in the presence of multiple user accesses
- Correct management of *replicas* (if they support it)

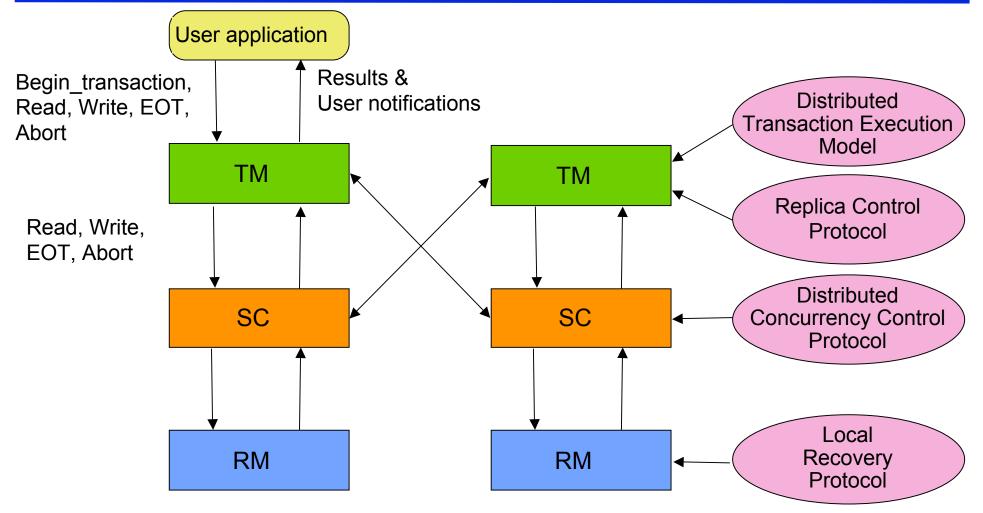
Architecture Revisited



Centralized Transaction Execution



Distributed Transaction Execution



Distributed DBMS

Concurrency Control

- 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.

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.

Global Non-serializability

T_1 : Read(x	$T_2: \operatorname{Read}(x)$	
$x \leftarrow x + \xi$	$x \leftarrow x*15$	
Write(2	\therefore Write(x)	
Commi	t Commit	

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

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

Concurrency Control Algorithms

Pessimistic

- Two-Phase Locking-based (2PL)
 - Centralized (primary site) 2PL
 - Primary copy 2PL
 - Distributed 2PL
- Timestamp Ordering (TO)
 - Basic TO
 - Multiversion TO
 - Conservative TO
- Hybrid
- Optimistic
 - Locking-based
 - Timestamp ordering-based

Locking-Based Algorithms

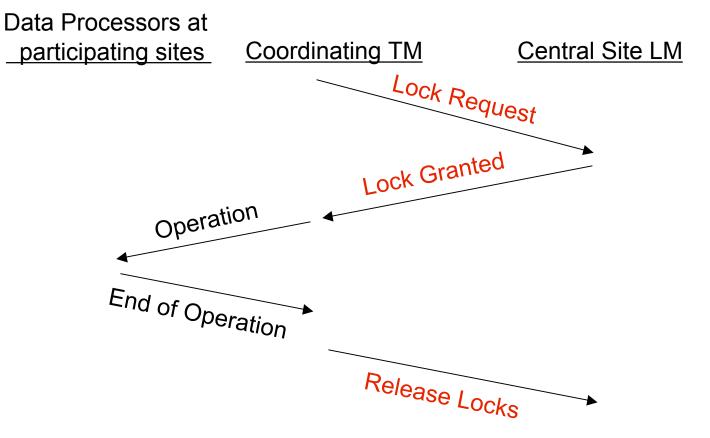
- Transactions indicate their intentions by requesting locks from the scheduler (called lock manager).
- Locks are either read lock (rl) [also called shared lock] or write lock (wl) [also called exclusive lock]
- Read locks and write locks conflict (because Read and Write operations are incompatible

	rl	wl
rl	yes	no
wl	no	no

Locking works nicely to allow concurrent processing of transactions.

Centralized 2PL

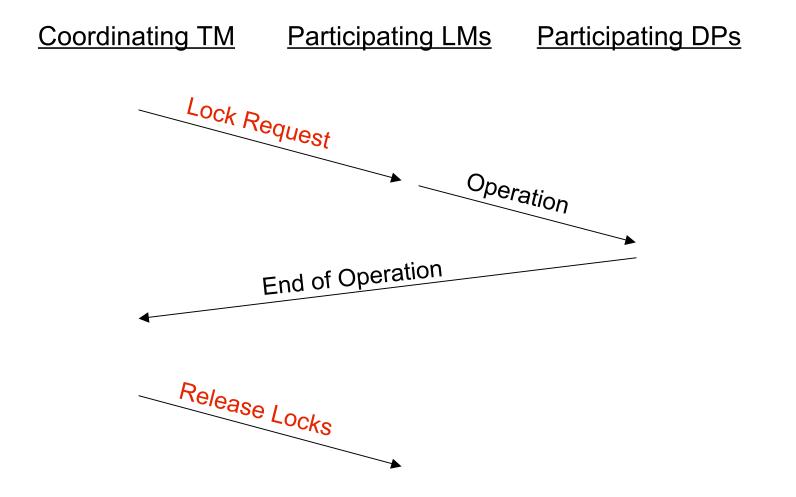
- There is only one 2PL scheduler in the distributed system.
- Lock requests are issued to the central scheduler.



Distributed 2PL

- 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.

Distributed 2PL Execution



Timestamp Ordering

- Transaction (T_i) is assigned a globally unique timestamp $ts(T_i)$.
- Pransaction 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
- Onflicting 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 $R_j(x)$ $rts(x) \leftarrow ts(T_j)$

for $W_i(x)$ **if** $ts(T_i) < rts(x)$ **and** $ts(T_i) < wts(x)$ **then** reject $W_i(x)$ **else** accept $W_i(x)$ $wts(x) \leftarrow ts(T_i)$

 $Distributed \ DBMS$

Outline

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



Problem:

How to maintain

atomicity

durability

properties of transactions

Distributed DBMS

Types of Failures

- Transaction failures
 - Transaction aborts (unilaterally or due to deadlock)
 - Avg. 3% of transactions abort abnormally
- System (site) failures
 - Failure of processor, main memory, power supply, ...
 - Main memory contents are lost, but secondary storage contents are safe
 - Partial vs. total failure
- Media failures
 - Failure of secondary storage devices such that the stored data is lost
 - Head crash/controller failure (?)
- Communication failures
 - Lost/undeliverable messages
 - Network partitioning

Distributed Reliability Protocols

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 \Rightarrow non-blocking termination

Two-Phase Commit (2PC)

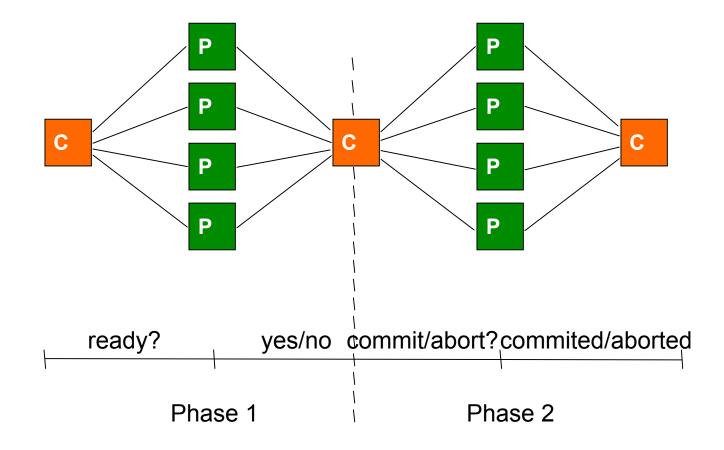
Phase 1 : The coordinator gets the participants ready to write the results into the database

- *Phase* 2 : Everybody writes the results into the database
 - Coordinator : The process at the site where the transaction originates and which controls the execution
 - Participant : The process at the other sites that participate in executing the transaction

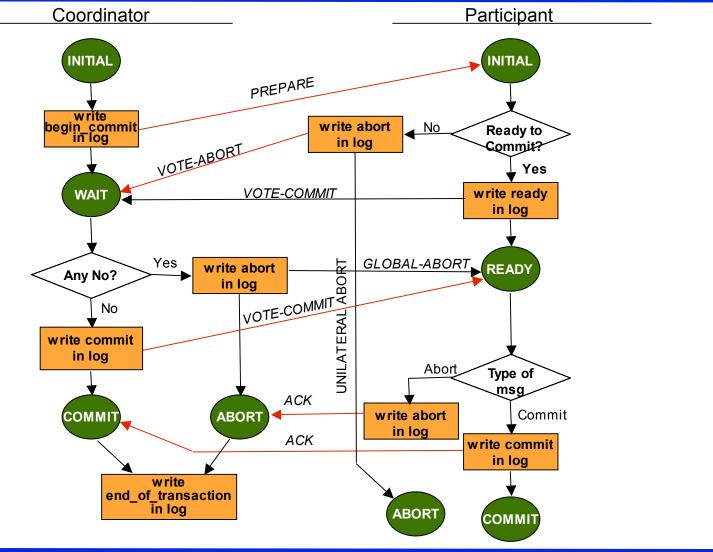
Global Commit Rule:

- The coordinator aborts a transaction if and only if at least one participant votes to abort it.
- Provide the coordinator commits a transaction if and only if all of the participants vote to commit it.

Centralized 2PC



2PC Protocol Actions



Distributed DBMS

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