An Overview of Query Optimization in Relational Systems

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Outline

- Preliminaries
- Query Optimization Framework
- Building Blocks
 - **■** Equivalence Transformations
 - Statistical Model
 - Tree-Finder: System-R, Volcano, Starburst
- Tuning Optimizers
- Beyond the Core Optimizer

5/14/9 Conclusion

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System R "Tree-Finder"

- Functionality
 - Ordering among joins
 - Chooses join methods and access paths
- Naïve strategy:
 - Generate all permutations of joins
 - Generate all combinations of join methods and access paths
 - Prohibitively expensive

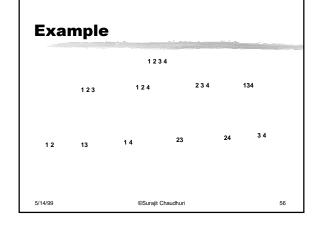
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Exploiting Dynamic Programming

- Best plan for Join(R,S) are the same for for
 - Join (Join (R,S), T)
 - Join (Join (R,S), V)
- Optimize one sub-expression once and reuse:
 - Join (Join (Join (R,S), T), V)
 - Join (Join (Join (R,S), V), T)
- One optimal plan for each subset of relations

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Enumeration Algorithm

- Find the optimal plan for $Join(T=\{R_1,...,R_n, R_{n+1}\})$
 - I For each Subset S (of size n) of $\{R_{1},..R_{n}, R_{n+1}\}$ do
 - I Let M = T S
 - I Find Optimal plan P(s) for Join(S)
 - I Determine optimal single-join plan for Join (P(s), M)
 - Iterate over choices of join methods and access methods

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- One optimal plan for each interesting order
- | Endfor
- I Pick the plan with the least cost

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Complexity

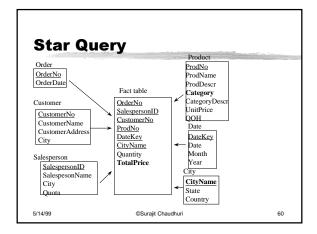
- Enumeration cost drops from O(n!) to O(n2^n)
- May need to store O(2^n) partial plans
 Do we?
- Significantly more efficient than the naïve scheme
- Significantly reduced number of "single-join" enumerations

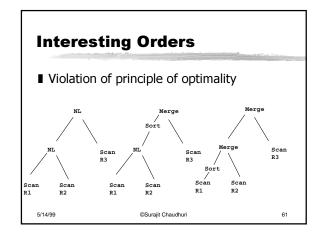
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Reducing Search in System-R Tree-Finder

- Avoid Cartesian product
 - I Defer all Cartesian products as late as possible to avoid "blow-up"
 - I Don't consider (R1 X R2) Join R3 if (R1 Join R3) Join R2 is feasible
 - May result in sub-optimality
 - I Large Sales Table
 - | Small Store Table with selection store.loc = "Redmond"
 - I Small Product Table with selection product.release = 1999
 - Recognized as "star" queries in OLAP

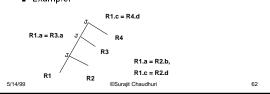
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Handling Interesting Orders in Tree-Finder

- Identify all columns that may exploit sorted order (by examining join predicates)
- Collapse into equivalent groups
- One optimal partial plan for each interesting order
- Example:



Key Ideas from System R

- Cost model
- Enumeration exploits
 - I Dynamic programming
 - I One optimal plan for each expression
 - I Violation of principle of optimality handled using interesting order
- Foundation of commercial optimizers

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Limitations of System R

- Limited Transformations
 - I Join ordering and choice of access methods only
 - Single block queries
- Not an adequate infrastructure for optimizing complex SQL

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Extensible Architectures

- Extensibility for optimizer developers
 - Add arbitrary transformations
 - Add new RE operators
- Generation of the operator tree is realized as a sequence of transformations
 - What sequence of transformations will result in a low-cost & valid RE operator tree?
- Example: Exodus/Volcano/Cascades, Starburst
- Lets try TEA (Toy Extensible Architecture)

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Plan Data Structure

- Data Structure representation of a query
 - Before Optimization: reflects query constructs (e.g., Join, Group By)
 - After Optimization: an operator tree that relational engine can execute (e.g., merge, sort)
 - During Optimization
 - I Some logical operators (RA)
 - I Some physical operators (RE)

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Life Cycle of a Plan in TEA

Join
Scan A

Join
Scan C

Scan B

Join
Scan C

Scan B

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Node Properties

- Logical
 - RA operator
 - Expression (sub-tree)
- Physical
 - ordering of rows
- **■** Estimated
 - cost, total cost (sub-tree)
 - statistical descriptors

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Transformation

- Condition-Action rules that
 - Preserve logical equivalence (Logical)
 - I RE operator to realize a logical expression (Implementation)
 - RE operator to realize a physical property (Enforcer)
- Bind template nodes to plan nodes
 - Verify condition
 - Apply action and generate a plan

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Naïve Tree-Finder

- Find (Node, Physical_Property, Cost-Limit)
 - Apply logical transformations to generate logically equivalent tree with root node' | Find (node', Physical_Property, Cost-Limit)
 - Apply implementation rule to a node
 - Find(child1, New1 Physical Property, cost-limit1)
 - 1 Find(child2, New2_Physical_Property, cost-limit2)
 - | Cost = Cost1 + Cost2 + Operator cost (stat1,stat2)
 - Enforce a physical property

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Efficiency Issues in TEA (1)

- Equivalence Classes
 - Expressions obtained by logical transformations
 - Lookup if an optimized plan exists for the class (using a hash table) or "in progress"
 - I Reusa
- Use branch and bound to drive the cost limit
 - Greedy algorithms

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Efficiency Issues in TEA (2)

- Rank applicable transformations
 - By promise (how?)
 - I Use implementation rules only for most promising logical expression
- Top-Down Optimizer
 - Redundant optimization of sub-expressions avoided
 - Memo structure with expression (history of transformations)

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Volcano v.s. Starburst

- Starburst
 - I Heuristic application of logical transformation
 - Cost-based mapping to RE operator trees
 - I Choice of access methods and join ordering for ASPJ queries
- Volcano
 - Uniformly cost-based
 - Harder to do search control

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Tuning Optimizer

- Information on the plan chosen by the optimizer
 - I Showplan (MS), Visual Explain (IBM) Interfaces
 - I Store plan information in tables
- Optimization Level
 - How exhaustive is the search for the "optimal" plan? I IBM DB2: greedy v.s. DP join enumeration
- Statistics
 - Create/Update Statistics
 - Manual update to statistics

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Tuning Optimizer: Hints

- Hints give partial control of execution back to the application developer
- Can specify
 - Join ordering, Join methods, Choice of Indexes
 - **I** Example

Select emp-id From Emp (index = 0) Where hire-date > `10/1/94'

■ Liability: Hints may result in poor performance with upgrades

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- Beyond the Core Optimizer
 - I Parallel and Distributed Systems
 - First/Top-K Queries
 - OLAP, Materialized Views
 - Semantic Query Optimization
 - Expensive Predicates, O-R Systems, Client-Server

■ Conclusion

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Distributed Systems

- Optimization in Distributed Systems
 - Site Selection for operations
 - Communication cost v.s. local processing time I Economic Model? (Cohera)
- Evolution of Distributed Systems
 - Scalability concerns
- => Parallel systems

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- Distributed information
- => Replicated sites

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Parallel Database Systems

- Objective is to minimize response time
- Forms of parallelism
 - I Independent, Pipelined, Partitioned
- Issues
 - I Consider Communication cost due to repartitioning
 - Scheduling of operators becomes an important aspect of optimization
 - ${\rm I\hspace{-.1em}I}$ Can/Should scheduling be separated from the rest of the query optimization?

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Parallel Database Systems

- Two step approach:
 - I Generate a sequential plan
 - Apply a scheduling algorithm to "parallelize" the plan
- The first phase should take into account cost of communication (e.g., repartitioning cost)
 - Influences partitioning attribute
- Scheduling algorithm assigns processors to operators
 - I Symmetric schedule: assigns each operator equally to each processor
 - I suboptimal when communication costs are considered

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First/Top K Queries

- Optimize query response
 - Produce first matching record quickly
- Top k restaurants in Seattle Order by customer-satisfaction
- Optimal query plan may be different
 - Use nested loop instead of sort-merge
 - Use non-clustered index scan instead of sort
- Commercial database systems provide constructs

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OLAP

- Spreadsheet paradigm drives the querying model
 - Backends always cannot digest complex SQL
 - Middleware ("ROLAP") tool optimizes SQL generation
 - I Creates and maintains materialized views
 - I Defines appropriate temporary relations

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Materialized Views

■ View Definitions

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- I Aggregation as part of view definitions
- Store Sales of Products by Quarter
- Optimization Problem
 - Materialized views enable additional logical transformations
 - I Check applicability of materialized views
 - Use a more specific view that can answer the query
 Sales of Products by Year is adequate to answer query
 on yearly revenue
- Need for a cost-based choice
 - I Multiple materialized views may apply
 - I Using base table may be better than using cached results ©Surajit Chaudhuri

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Semantic Optimization

- Transformation Rules in Classical Optimizer
 - Equivalences over SQL
- Semantic Optimization imports application knowledge
 - ${\rm I\hspace{-.1em}I} \hspace{.1em} \hbox{Constraints that hold over a database}$
 - I No person above the age of 25
 - Optimizer can exploit these constraints

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Expensive Predicates, O-R Systems

- Expensive Predicates: Cannot push down selections
 - Select * From Stocks Where stocks. fluctuation > .6
 - Associate a per-tuple CPU and IO cost with predicate evaluation
- O-R Systems: Relationship among udfs
 - Spatial data-blade may support related spatial indexes
 - I Use rules to specify semantic relationships
 - $\hbox{{\it I}} \quad \hbox{Cost-based semantic Query Optimization}$
 - I New issues in costing and enumeration
 - I How to use costs uniformly across ADT-s
 - I "Mix and match" or "ADT-specific" optimization?

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Other Issues

- Compile Time v.s. Run time optimization
 - Choose plan and Exchange
- Resource governor
 - \blacksquare Adapting optimization to memory constraints
- Sensitivity of the cost model
- Language Extension for ease of optimization (e.g., Cube)
- Client-Server issues
 - Function shipping, Data shipping or mixed mode

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Concluding Remarks

- Quality of the Optimizer depends on
 - Cost Estimator, Transformation Rules, Search Control/Enumerator
- Many external factors influence performance
 - Query Processing engine
 Physical database design
- Oversimplification may render results useless
 - Need to pay attention to SQL semantics
- Questions?
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