Distributed Queries

```
SELECT AVG(E.age)
FROM Employees E
WHERE E.salary > 3000
AND E.salary < 7000</pre>
```

Horizontally Fragmented:

Tuples with Salary < 5000 at London, >= 5000 at Paris.

- Must compute SUM(*age*), COUNT(*age*) at both sites.
- If WHERE contained just E.salary>6000, just one site.
- Vertically Fragmented: *title* and *salary* at London, *ename* and *age* at Paris, *id* at both.
 - Must reconstruct relation by join on *id*, then evaluate the query.
- Replicated: Employees copies at both sites.
 - Choice of site based on local costs, shipping costs.

Distributed Joins



- Ship to One Site: Ship Reports to London.
 - Cost: 1000 S + 4500 D (sort-merge join; cost = 3*(500+1000))
 - If result size is very large, may be better to ship both relations to result site and then join them!
- Ship Employees to Paris
 - Cost: 500 S + 4500 D

Distributed Joins

for Employees 1 page = 80 tuples for Reports 1 page = 100 tuples



PARIS
Reports

- Fetch as Needed, Page oriented nested loops, Employees as outer (for each Employee page fetch all Reports pages from Paris):
 - Cost: 500 D + 500 * 1000 (D+S)
 - **D** is cost to read/write page; **S** is cost to ship page.
 - If query was not submitted at London, must add cost of shipping result to query site.
 - Can also do INL (indexed nested loops) at London, fetching matching Reports tuples to London as needed.

Semijoin

- At London, project Employees onto join columns and ship this to Paris.
- At Paris, join Employees projection with Reports.
 - Result is called **reduction** of Reports with respect to Employees .
- Ship reduction of Reports to London.
- At London, join Employees with reduction of Reports.
- Idea: Tradeoff the cost of computing and shipping projection and computing and shipping reduction for cost of shipping full Reports relation.
- Especially useful if there is a selection on Employees, and answer desired at London.

Bloomjoin

- At London, compute a bit-vector of some size k:
 - Hash join column values into range 0 to k-1.
 - If some tuple hashes to i, set bit i to 1 (i from 0 to k-1).
 - Ship bit-vector to Paris.
- At Paris, hash each tuple of Reports similarly, and discard tuples that hash to 0 in Employees bitvector.
 - Result is called **reduction** of Reports wrt Employees.
- Ship bit-vector reduced Reports to London.
- At London, join Employees with reduced Reports.
- Bit-vector cheaper to ship, almost as effective.

Distributed Query Optimization

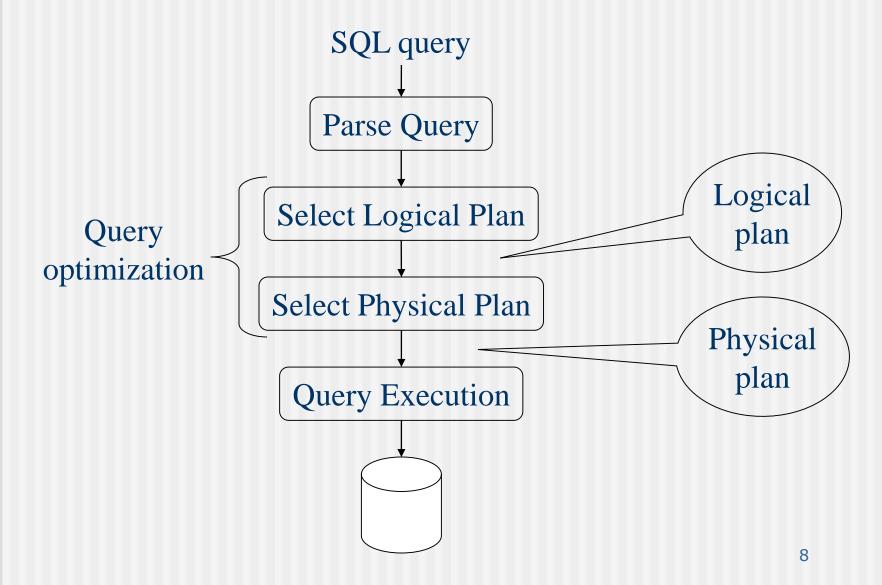
- Cost-based approach; consider all plans, pick cheapest; similar to centralized optimization.
 - Difference 1: Communication costs must be considered.
 - Difference 2: Local site autonomy must be respected.
 - Difference 3: New distributed join methods.

- Query site constructs global plan, with suggested local plans describing processing at each site.
 - If a site can improve suggested local plan, free to do so.

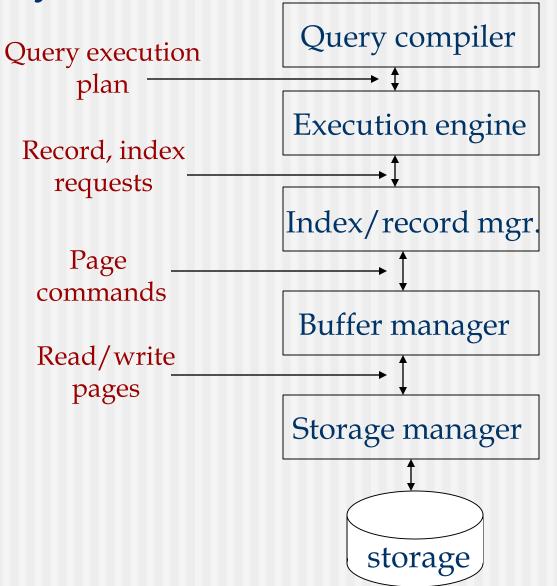
COURSE 10

Query Optimization

Architecture of Database Engine



Query Execution



Schema for Examples

```
Students (sid: integer, sname: string, age: integer)
Courses (cid: integer, name: string, location: string)
Evaluations (sid: integer, cid: integer, day: date, grade: integer)
```

■ Students:

■ Each tuple is 50 bytes long, 80 tuples per page, 500 pages.

Courses:

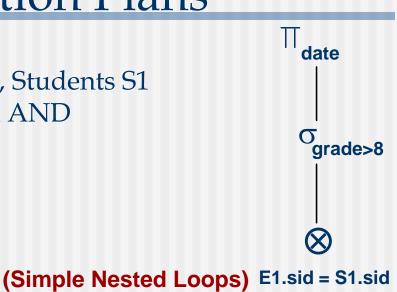
■ Each tuple is 40 bytes long, 100 tuples per page, 1 page.

■ Evaluations:

■ Each tuple is 40 bytes long, 100 tuples per page, 1000 pages.

Query Execution Plans

SELECT E1.date
FROM Evaluations E1, Students S1
WHERE E1.sid=S1.sid AND
E1.grade > 8



Query Plan:

- logical tree
- implementation choice at every node
- scheduling of operations.

Evaluations (E1) Students (S1)
(Table scan) (Index scan)

Some operators are from relational algebra, and others (e.g., scan, group) are not.

The Leaves of the Plan: Scans

- Table scan: iterate through the records of the relation.
- Index scan: go to the index, from there get the records in the file
- Sorted scan: produce the relation in order. Implementation depends on relation size.
- How do we combine Operations?
 - The iterator model. Each operation is implemented by 3 functions:
 - *Open*: sets up the data structures and performs initializations
 - *GetNext*: returns the next tuple of the result.
 - Close: ends the operations. Cleans up the data structures.
 - Enables pipelining!
 - Contrast with data-driven materialize model.
 - Sometimes it's the same (e.g., sorted scan).

Query Optimization Process

- Parse the SQL query into a logical tree:
 - identify distinct blocks (corresponding to nested subqueries or views).
- Query rewrite phase:
 - apply algebraic transformations to yield a cheaper plan.
 - Merge blocks and move predicates between blocks.
- Optimize each block: join ordering.
- Complete the optimization: select scheduling (pipelining strategy).

Overview of Query Optimization

- *Plan*: Tree of R.A. ops, with choice of alg for each op.
 - Each operator typically implemented using a `pull' interface: when an operator is `pulled' for the next output tuples, it `pulls' on its inputs and computes them.
- Two main issues:
 - For a given query, what plans are considered?
 - Algorithm to search plan space for cheapest (estimated) plan.
 - How is the cost of a plan estimated?
- Ideally: Want to find best plan.
- Practically: Avoid worst plans!

Highlights of System R Optimizer

- Impact:
 - Most widely used currently; works well for < 10 joins.
- Cost estimation: Approximate art at best.
 - Statistics, maintained in system catalogs, used to estimate cost of operations and result sizes.
 - Considers combination of CPU and I/O costs.
- Plan Space: Too large, must be reduced.
 - Only the space of *left-deep plans* is considered.
 - Left-deep plans allow output of each operator to be <u>pipelined</u> into the next operator without storing it in a temporary relation.
 - Cartesian products avoided.

Motivating Example

SELECT S.sname
FROM Evaluations E, Students S
WHERE E.sid=S.sid AND
E.cid=100 AND S.age>22

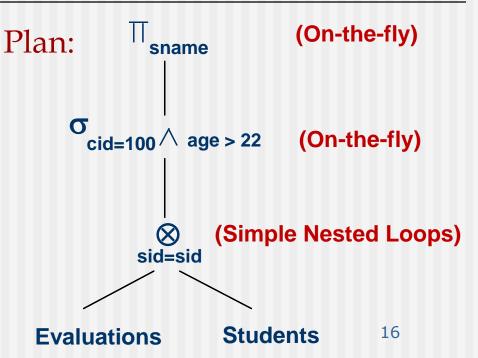
RA Tree: sname

Ocid=100 age > 22

Sid=sid

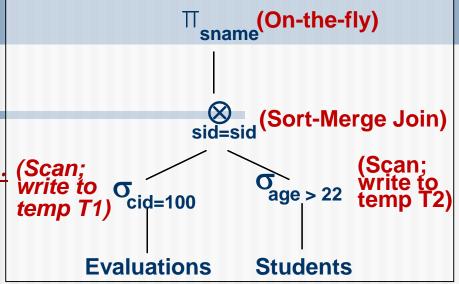
Evaluations Students

- Cost: 500+500*1000 I/Os
- By no means the worst plan!
- Misses several opportunities: selections could have been `pushed' earlier, no use is made of any available indexes, etc.
- *Goal of optimization:* To find more efficient plans that compute the same answer.



Alternative Plans 1

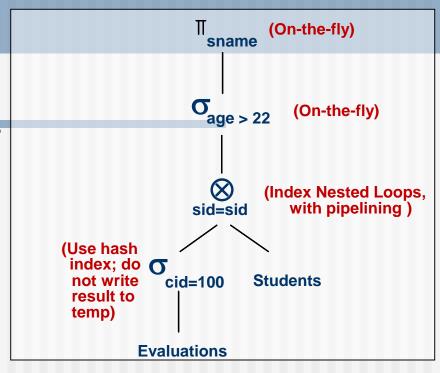
- Main difference: push selects. (Scan; write to
- With 5 buffer pages,
- cost of plan:
 - Scan Evaluations (1000) + write temp T1 (10 pages, if we have 100 courses, uniform distribution). *total* 1010 I/Os
 - Scan Students (500) + write temp T2 (250 pages, if we have 10 ages). *total 750 I/Os*
 - Sort T1 (2*2*10), sort T2 (2*3*250), merge (10+250) *total 1800 I/Os*
 - Total: 3560 page I/Os.
- If we used BNL join, join cost = 10+4*250, total cost = 2770
- If we `push' projections, T1 has only *sid*, T2 only *sid* and *sname*:
 - T1 fits in 3 pages, cost of BNL drops to under 250 pgs, total < 2000.



Alternative Plans 2

- With clustered index on *cid* of Evaluations, we get 100,000/100 = 1000 tuples on 1000/100 = 10 pages.
- INL with *pipelining* (outer is not materialized).
 - Projecting out unnecessary fields from outer doesn't help.
- Join column *sid* is a key for Students.
 - At most one matching tuple, un-clustered index on *sid* OK.
- Decision not to push age>22 before the join is based on availability of sid index on Students.

Cost: Selection of Evaluations tuples (10 I/Os); for each, must get matching Students tuple (1000*1.2); total 1210 I/Os.



Query Blocks: Units of Optimization

- An SQL query is parsed into a collection of *query blocks*, and these are optimized one block at a time.
- Nested blocks are usually treated as calls to a subroutine, made once per outer tuple. (This is an oversimplification, but serves for now.)

FROM Students S
WHERE S.age IN
(SELECT MAX (S2.age)
FROM Students S2
GROUP BY S2.sname)

Outer block Nest

Nested block

For each block, the plans considered are:

- All available access methods, for each reln in FROM clause.
- All *left-deep join trees* (i.e., all ways to join the relations one-at-a-time, with the inner reln in the FROM clause, considering all reln permutations and join methods.)

Cost Estimation

- For each plan considered, must estimate cost:
 - Must estimate *cost* of each operation in plan tree.
 - Depends on input cardinalities.
 - We've already discussed how to estimate the cost of operations (sequential scan, index scan, joins, etc.)
 - Must estimate *size of result* for each operation in tree!
 - Use information about the input relations.
 - For selections and joins, assume independence of predicates.
- We'll discuss the System R cost estimation approach.
 - Very inexact, but works ok in practice.
 - More sophisticated techniques known now.

Relational Algebra Equivalences

- Allow us to choose different join orders and to `push' selections and projections ahead of joins.
- <u>Selections</u>: $\sigma_{c1\wedge...\wedge cn}(R) \equiv \sigma_{c1}(...(\sigma_{cn}(R)))$ (Cascade) $\sigma_{c1}(\sigma_{c2}(R)) \equiv \sigma_{c2}(\sigma_{c1}(R))$ (Commute)
- <u>Projections</u>: $\pi_{a1}(R) \equiv \pi_{a1}(\dots(\pi_{an}(R)))$ (Cascade)
- <u>Joins</u>: $R \otimes (S \otimes T) \equiv (R \otimes S) \otimes T$ (Associative) $(R \otimes S) \equiv (S \otimes R)$ (Commute)
- + show that: $R \otimes (S \otimes T) \equiv (T \otimes R) \otimes S$

More Equivalences

- A projection commutes with a selection that only uses attributes retained by the projection.
- Selection between attributes of the two arguments of a cross-product converts cross-product to a join.
- A selection on just attributes of R commutes with $R \otimes S$. (i.e., $\sigma (R \otimes S) \equiv \sigma (R) \otimes S$)
- Similarly, if a projection follows a join $\mathbb{R} \otimes \mathbb{S}$, we can `push' it by retaining only attributes of \mathbb{R} (and \mathbb{S}) that are needed for the join or are kept by the projection.

Enumeration of Alternative Plans

- There are two main cases:
 - Single-relation plans
 - Multiple-relation plans
- For queries over a single relation, queries consist of a combination of selects, projects and aggregate ops:
 - Each available access path (file scan / index) is considered, and the one with the least estimated cost is chosen.
 - The different ops are essentially carried out together (e.g., if an index is used for a selection, projection is done for each retrieved tuple, and the resulting tuples are *pipelined* into the aggregate computation).

Cost Estimates for Single-Relation Plans

- Index I on primary key matches selection:
 - Cost is Height(I)+1 for a B+ tree, about 1.2 for hash index.
- Clustered index I matching one or more selects:
 - \blacksquare (NPages(I)+NPages(R)) * product of RF's of matching selects.
- Non-clustered index I matching one or more selects:
 - \blacksquare (NPages(I)+NTuples(R)) * product of RF's of matching sels.
- Sequential scan of file:
 - \blacksquare *NPages(R).*
- + **Note**: Typically, no duplicate elimination on projections! (Exception: Done on answers if user says DISTINCT.)

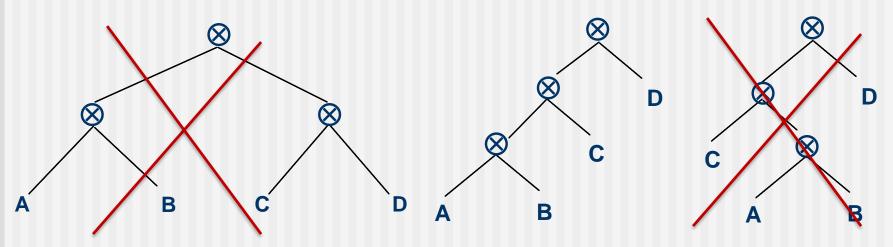
Example

SELECT S.sid FROM Students S WHERE S.age=20

- If we have an index on *age*:
 - \blacksquare (1/NKeys(I)) * NTuples(R) = (1/10) * 40000 tuples retrieved.
 - Clustered index: (1/NKeys(I)) * (NPages(I)+NPages(R)) = (1/10) * (50+500) pages are retrieved. (This is the *cost*)
 - Unclustered index: (1/NKeys(I)) * (NPages(I)+NTuples(R)) = (1/10) * (50+40000) pages are retrieved.
- If we have an index on *sid*:
 - Would have to retrieve all tuples/pages. With a clustered index, the cost is 50+500, with un-clustered index, 50+40000.
- Doing a file scan:
 - We retrieve all file pages (500).

- Queries Over Multiple Relations

 Fundamental decision in System R: only left-deep join trees are considered.
 - As the number of joins increases, the number of alternative plans grows rapidly; we need to restrict the search space.
 - Left-deep trees allow us to generate all *fully pipelined* plans.
 - -Intermediate results not written to temporary files.
 - -Not all left-deep trees are fully pipelined (e.g., SM join).



Enumeration of Left-Deep Plans

- Left-deep plans differ only in the order of relations, the access method for each relation and the join method for each join.
- Enumerated using N passes (if N relations joined):
 - Pass 1: Find best 1-relation plan for each relation.
 - Pass 2: Find best way to join result of each 1-relation plan (as outer) to another relation. (All 2-relation plans.)
 - Pass N: Find best way to join result of a (N-1)-relation plan (as outer) to the N'th relation. (All N-relation plans.)
- For each subset of relations, retain only:
 - Cheapest plan overall, plus
 - Cheapest plan for each *interesting order* of the tuples.

Enumeration of Plans (cont.)

- ORDER BY, GROUP BY, aggregates etc. handled as a final step, using either an `interestingly ordered' plan or an additional sorting operator.
- An N-1 way plan is not combined with an additional relation unless there is a join condition between them, unless all predicates in WHERE have been used up.
 - i.e., avoid Cartesian products if possible.
- In spite of pruning plan space, this approach is still exponential in the number of tables.

Example

Pass1:

Students:
B+ tree on age
Hash on sid
Evaluations:
B+ tree on cid

■ *Students*: B+ tree matches *age*>22, and is probably cheapest. However, if this selection is expected to retrieve a lot of tuples, and index is un-clustered, file scan may be cheaper.

- Still, B+ tree plan kept (because tuples are in age order).
- *Evaluations*: B+ tree on *cid* matches *cid*=100; cheapest.

Pass 2:

We consider each plan retained from Pass 1 as the outer, and consider how to join it with the (only) other relation. (e.g., *Evaluations as outer*: Hash index can be used to get Students tuples that satisfy *sid* = outer tuple's *sid* value.)

Nested Queries

- Nested block is optimized independently, with the outer tuple considered as providing a selection condition.
- Outer block is optimized with the cost of `calling' nested block computation taken into account.
- Implicit ordering of these blocks means that some good strategies are not considered.
 The non-nested version of the query is typically optimized better.

SELECT S.sname FROM Students S WHERE EXISTS (SELECT * FROM Evaluations E WHERE E.cid=103 AND E.sid=S.sid)

Nested block to optimize:

SELECT *

FROM Evaluations E

WHERE E.cid=103

AND S.sid= outer value

Equivalent non-nested query:
SELECT S.sname
FROM Sudents S,
Evaluations E
WHERE S.sid=E.sid
AND E.cid=103