

Pipelining

- Parent and child execute concurrently
- Parent and child share buffer space
 - k-block shared (sub)buffer
 - child produces k blocks of output Fill buffer
 - parent consumes k blocks of input from child –
 - Empty buffer – NO disk write cost child;
 - NO disk read cost parent
- Algorithms of child and parent must support this

 Child: usually does; produce 1 block output at a time
 - Parent: choice of algorithm critical !

Algorithms for parent - JOIN

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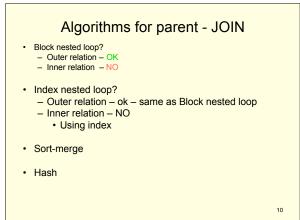
- Block nested loop?
- · Index nested loop?
- Sort-merge
- · Hash

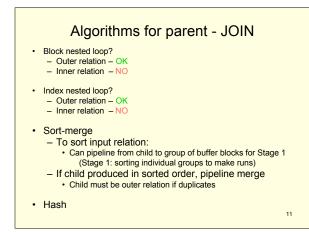
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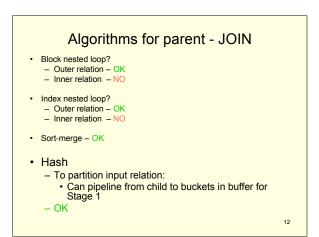
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Algorithms for parent - JOIN

- · Block nested loop?
 - Outer relation ok
 - Read relation once, "chunk" by "chunk"
 - Use shared buffer for "chunk"
 - Inner relation NO
 - Must re-read entire inner relation for every "chunk" of outer
- Index nested loop?
- · Sort-merge
- Hash







Allocating buffer blocks

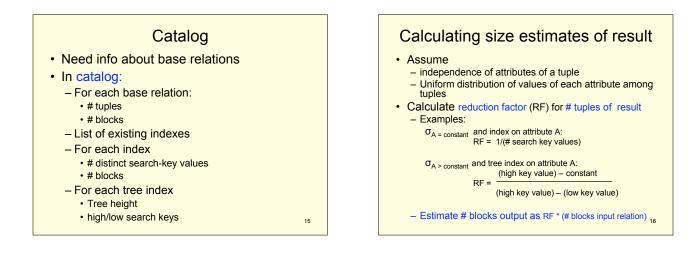
- If have simultaneous pipelining up tree

 How many buffer blocks for each child-to
 - parent exchange?
 Affects speed of algorithms
- Limit number of simultaneous pipelines
- If no pipeline between child and parent materialize result of child
 - Child writes result to disk
 - Parent reads from disk

Multi-operation query

- Want plan
 - Parse tree
 - Pipelining plan for each edge
 - Algorithm for each interior node (operation)
- To build plan
 - Consider alternatives
 - ALL?
 - Estimate costs
 Choose "best"
 - Really "good enough"

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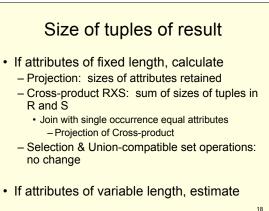
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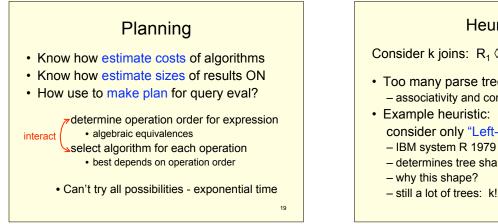
Reduction factor of joins

 Estimate # tuples of (R◊◊S) on shared attribute A as

RF * (# tuples R) * (# tuples S) - Looking at join as selection on RXS

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Heuristics

Consider k joins: $R_1 \diamond \diamond R_2 \diamond \diamond \dots \diamond \diamond R_k$

- · Too many parse trees - associativity and commutativity
- Example heuristic: consider only "Left-deep join trees"

 - determines tree shape, not order R_i

 - still a lot of trees: k!

Algorithm design

• Observe for $(R_1 \diamond \diamond R_2 \diamond \diamond \dots R_{k-1}) \diamond \diamond R_k$:

- once decide least-cost way do () actual order compute w/in () not affect best choice for () $\Diamond \Diamond R_{\mu}$
- whether () result sorted or hashed does affect best choice for () $\Diamond \Diamond R_k$
- ⇒dynamic programming algorithm
 - · walk up left-deep tree

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Using dynamic programming

For node distance d from leftmost leaf,

- estimate lowest cost of evaluating subtree for • each size-(d+1) subset of {R_i} 1. without regard to order of result records 2. in each "natural" sorted order of result records
- Use results from child node
- Include pipelining strategy
- Remember best plans and pipelining strategy for each subset
 - can reconstruct order going back down tree
- Running time exponential in k still consider each subset of {R_i}
 - don't consider each order of R_i's at next level

Other operations

- · Move selects and projects up/down tree
- Try to push selects down tree Pushing projects can also be useful - whv?
 - not always good idea: destroys indexes
- · can include in left-join-tree analysis
- · Text has detailed discussion equivalences for relational algebra operations

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Index-only Algorithms

If have indexes giving pointers to records for all relations in query, consider:

- Use indexes to execute operations • must have right search keys
- Retrieve records only at end
- If need only count, never retrieve full records

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Summary

- Have seen in detail how to execute joins
- Have considered execution of other relational alg. op.s
- · Have looked at how estimate sizes of results
- Have briefly considered one heuristic for making plan for several joins
 - restrict to left-deep trees
- Have looked briefly at planning when relational alg. expr. has more than just joins

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