

## Announcements (March 1)

- \* Reading assignment due Wednesday
  - Buffer management
- Homework #2 due this Thursday
- \* Course project proposal due in one week
- \* Midterm next Thursday in class
  - Open book, open notes

# Announcements (March 3)

- \* No more reading assignment before midterm
- Homework #2 due today
- Will be graded by next Tuesday
- Midterm next Thursday in class
  - Open book, open notes
  - Everything up to (and including) today's lecture
  - Format similar to sample midterm from last year (available only in hardcopies; solution to be handed out next Tuesday), but shorter <sup>©</sup>
- \* Course project proposal due next Tuesday

# Physical (execution) plan

- A complex query may involve multiple tables and various query processing processing algorithms
  - E.g., table scan, index nested-loop join, sort-merge join, hash-based duplicate elimination...
- ♦ A physical plan for a query tells the DBMS query processor how to execute the query
  - A tree of physical plan operators
  - Each operator implements a query processing algorithm
  - Each operator accepts a number of input tables/streams and produces a single output table/stream





• Children pipeline their results to parents

# Iterator interface

Every physical operator maintains its own execution state and implements the following methods:

- open(): Initialize state and get ready for processing
- getNext(): Return the next tuple in the result (or a null pointer if there are no more tuples); adjust state to allow subsequent tuples to be obtained
- close(): Clean up

# An iterator for table scan

## \$ open()

Allocate a block of memory

## \$ getNext()

- If no block of *R* has been read yet, read the first block from the disk and return the first tuple in the block (or the null pointer if *R* is empty)
- If there is no more tuple left in the current block, read the next block of *R* from the disk and return the first tuple in the block (or the null pointer if there are no more blocks in *R*)
- Otherwise, return the next tuple in the memory block

## \$ close()

Deallocate the block of memory



# An iterator for 2-pass merge sort

## \$ open()

- Allocate a number of memory blocks for sorting
- Call open() on child iterator
- getNext()
  - If called for the first time
    - Call getNext() on child to fill all blocks, sort the tuples, and output a run
      Repeat until getNext() on child returns null
    - Read one block from each run into memory, and initialize pointers to point to the beginning tuple of each block
  - Return the smallest tuple and advance the corresponding pointer; if a block is exhausted bring in the next block in the same run

#### close()

- Call close() on child
- Deallocate sorting memory and delete temporary runs



# <sup>12</sup> Execution of an iterator tree Call root.open() Call root.getNext() repeatedly until it returns null Call root.close() Requests go down the tree Intermediate result tuples go up the tree No intermediate files are needed But maybe useful if an iterator is opened many times

 Example: complex inner iterator is opened many times
 Example: complex inner iterator tree in a nested-loop join; "cache" its result in an intermediate file

# Memory management for DBMS

- DBMS operations require main memory
  - While data resides on disk, it is manipulated in memory
  - Sometimes the more memory the better, e.g., sort
- One approach: let each operation pre-allocate some amount of "private" memory and manage it explicitly
  - Not very flexible
  - Limits sharing and reuse
- Alternative approach: use a buffer manager
  - Responsible for reading/writing data blocks from/to disk as needed
  - Higher-level code can be written without worrying about whether data is in memory or not

# Buffer manager basics

- \* Higher-level code can pin and unpin a frame
  - Pin: I need to work on this frame in memory
  - Unpin: I no longer need this frame
  - A completely unpinned frame is a candidate for replacement
     In some systems you can hate a frame (i.e., suggesting it for replacement)
- \* A frame becomes dirty when it is modified
  - Only dirty frames need to be written back to disk
  - @Related to transaction processing

# Standard OS replacement policies

- Example
  - Current buffer pool: 0, 1, 2
  - Past requests: 0, 1, 2
  - Incoming requests: 3, 0, 1, 2, 3, 0, 1, 2, 3, 4, 5, 6, 7, ...
     Which frame to replace?
- Optimal: replace the frame that will not be used for the longest time (2)
- Random (0, 1, or 2 with equal probability)
- LRU: least recently used (0)
- LRU approximation: clock, aging
- MRU: most recently used (2)

# Problems with OS buffer management

Stonebraker. "Operating System Support for Database Management." CACM, 1981.

#### Performance problems

- Getting a page from the OS to user space is usually a system call (process switch) and copy
- Replacement policy
  - LRU, clock, etc. often ineffective
  - DBMS knows access pattern in advance and therefore should dictate policy → major OS/DBMS distinction
- \* Prefetch policy
  - DBMS knows of multiple "orders" for a set of records; OS only knows physical order
- Crash recovery
  - DBMS needs more control

## Next

Chou and DeWitt. "An Evaluation of Buffer Management Strategies for Relational Database Systems." VLDB 1985.

- \* Old algorithms
  - Domain separation algorithm
  - "New" algorithm
  - Hot set algorithm
- \* Query locality set model
- \* DBMIN algorithm

# Domain separation algorithm

- Split work/memory into domains; LRU within each domain; borrow from other domains when out of frames
  - Example: one domain for each level of the B<sup>+</sup>-tree

#### Limitations

- Assignment of pages to domains is static, and ignores how pages are used
  - Example: A data page is accessed only once in a scan, but the same data page is accessed many times in a NLJ
- Does not differentiate relative importance between types of pages
   Example: An index page is more important than a data page
- Memory allocation is based on data rather queries → need orthogonal load control to prevent thrashing

# The "new" algorithm

- To Observations based on the reference patterns of queries
  - Priority is not a property of a data page, but of a relation
  - Each relation needs a "working set"
- \* Divide buffer pool into chunks, one per relation
- Prioritize relations according to how often their pages are reused
- Replace a frame from the least reused relation and add it to the chunk of the referenced relation
- \* Each active relation is guaranteed with one frame
- MRU within each chunk (seems arbitrary)
- \* Simulations look good; implementation did not beat LRU

# Hot set algorithm

- Texploit query behavior more!
- A set of pages that are accessed over and over form a hot set
   "Hot points" in the graph of buffer size vs. number of page faults
  - Example: For nested-loop join  $R \bowtie S$ , size of hot set is B(S) + 1 (under LRU)
- \* Each query is given enough memory for its hot set
- Admission control: Do not let a query into the system unless its hot set fits in memory
- \* Replacement: LRU within each hot set (seems arbitrary)
- Derivation of hot set assumes LRU, which may be suboptimal
  - Example: What is better for nested-loop join?

# Query locality set model

- \* Observations
  - DBMS supports a limited set of operations
  - Reference patterns are regular and predictable
  - Reference patterns can be decomposed into simple patterns
- Reference pattern classification
  - Sequential
  - Random
  - Hierarchical

# Sequential reference patterns

- $\boldsymbol{\diamond}$  Straight sequential: read something sequentially once
  - Example: selection on unordered table
  - Feach page is only touched once, so just buffer one page
- $\boldsymbol{\ast}$  Clustered sequential: repeatedly read a "chunk" sequentially

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- Example: merge join; rows with the same join column value are scanned multiple times
- "Keep all pages in the chunk in buffer
- Looping sequential: repeatedly read something sequentially
   Example: nested-loop join
  - "Keep as many pages as possible in buffer, with MRU replacement

## Random reference patterns

- Independent random: truly random accesses
  - Example: index scan through a non-clustered (e.g., secondary) index yields random data page access
  - The larger the buffer the better?
- Clustered random: random accesses that happen to demonstrate some locality
  - Example: in an index nested-loop join, inner index is non-clustered and non-unique, while outer table is clustered and non-unique
  - Try to keep in buffer data pages of the inner table accessed in one cluster

# Hierarchical reference patterns

- Example: operations on tree indexes
- Straight hierarchical: regular root-to-leaf traversal
- Hierarchical with straight sequential: traversal followed by straight sequential on leaves
- Hierarchical with clustered sequential: traversal followed by clustered sequential on leaves
- Looping hierarchical: repeatedly traverse an index
  - Example: index nested-loop join
  - Keep the root index page in buffer

# DBMIN algorithm

- Associate a chunk of memory with each file instance (each table in FROM)
  - This chunk is called the file instance's locality set
  - Instances of the same table may share buffered pages
  - But each locality set has its own replacement policy
     Based on how query processing uses each relation (finally!)
     No single policy for all pages accessed by a query
     No single policy for all pages in a table
- Estimate locality set sizes by examining the query plan and database statistics
- Admission control: a query is allowed to run if its locality sets fit in free frames

# DBMIN algorithm (cont'd)

- \* Locality sets: each "owns" a set of pages, up to a limit l
- Global free list: set of "orphan" pages
- \* Global table: allow sharing among concurrent queries
- ♦ Query q requests page p
  - If p is in memory and in q's locality set
     Just update usage statistics of p
  - If p is in memory and in some other query's locality set
     Just make p available to q; no further action is required
  - If *p* is in memory and in the global free list
    - Add p to q's locality set; if q's locality set exceeds its size limit, replace a page (release it back to the global free list)
  - If p is not in memory
    - Use a page from global free list to get p in; proceed as in the previous case

# Locality sets for various ref. patterns

- \* Straight sequential
  - Size = 1
  - Just replace as needed
- Clustered sequential
  - Size = number of pages in the largest cluster
  - FIFO or LRU (assuming large enough size)
- \* Looping sequential
  - Size = number of pages in the table
  - MRU

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# Locality sets for more ref. patterns

 Straight hierarchical, hierarchical/straight sequential: just like straight sequential

- Size = 1
- Just replace as needed
- Hierarchical/clustered sequential: like clustered sequential
  - Size = number of index pages in the largest cluster
  - FIFO or LRU
- Looping hierarchical
  - At each level of the index you have random access among pages
  - Use Yao's formula to figure out how many pages need to be
  - accessed at each level
  - Size = sum over all levels that you choose to worry about
  - LIFO with 3-4 buffers should be okay

# Simulation study

- Hybrid simulation model
  - Trace-driven simulation
    - Recorded from a real system (running Wisconsin Benchmark)
    - For each query, record its execution trace
    - Page read/write, file open/close, etc.
  - Distribution-driven simulation
    - Generated by some stochastic model
    - Synthesize the workload by merging query execution traces
- Simulator models CPU, memory, and one disk
- Performance metric: query throughput















# Conclusion

- Same basic access patterns come up again and again in query processing
- \* Make buffer manager aware of these access patterns
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  - Contents can at best offer guesses at likely workloads