

Database Systems

(資料庫系統)

December 10/12, 2007

Lecture #11

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Announcement

- Assignment #5 is due next Monday (outside TA office 336/338).

2

External Sorting

Chapter 13

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Why learn sorting again?

- $O(n^2)$: bubble, insertion, selection, ... sorts
- $O(n \log n)$: heap, merge, quick, ... sorts
- Sorting huge dataset (say 10 GB)
- CPU time complexity may mean little on practical systems
- Why?

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“External” Sorting Defined

- Refer to sorting methods when the data is too large to fit in main memory.
 - E.g., sort 10 GB of data in 100 MB of main memory.
- During sorting, some intermediate steps may require data to be stored externally on disk.
- Disk I/O cost is much greater than CPU instruction cost
 - Average disk page I/O cost: 10 ms vs. 4 GHz CPU clock: 0.25 ns.
 - Minimize the disk I/Os (rather than number of comparisons).

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Outline (easy chapter)

- Why does a DMBS sort data?
- Simple 2-way merge sort
- Generalize B-way merge sort
- Optimization
 - Replacement sort
 - Blocked I/O optimization
 - Double buffering
- Using an existing B+ tree index vs. external sorting

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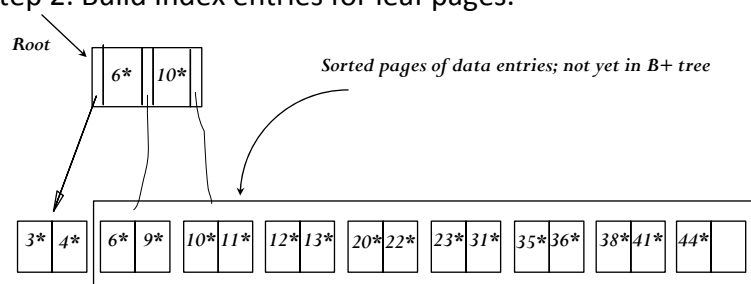
When does a DBMS sort data?

- Users may want answers to query in some order
 - E.g., students sorted by increasing age
- Sorting is the first step in bulk loading a B+ tree index
- Sorting is used for eliminating duplicate copies
- Join requires a sorting step.
 - Sort-join algorithm requires sorting.

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Bulk Loading of a B+ Tree

- Step 1: Sort data entries. Insert pointer to first (leaf) page in a new (root) page.
- Step 2: Build Index entries for leaf pages.



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Example of Sort-Merge Join

<i>sid</i>	<i>sname</i>	<i>rating</i>	<i>age</i>
22	dustin	7	45.0
28	yuppy	9	35.0
31	lubber	8	55.5
44	guppy	5	35.0
58	rusty	10	35.0

<i>sid</i>	<i>bid</i>	<i>day</i>	<i>rname</i>
28	103	12/4/96	guppy
28	103	11/3/96	yuppy
31	101	10/10/96	dustin
31	102	10/12/96	lubber
31	101	10/11/96	lubber
58	103	11/12/96	dustin

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A Simple Two-Way Merge Sort

- External sorting: data \gg memory size
- Say you only have 3 (memory) buffer pages.
 - You have 7 pages of data to sort.
 - Output is a sorted file of 7 pages.
- How would you do it?
- How would you do it if you have 4 buffer pages?
- How would you do it if you have n buffer pages?

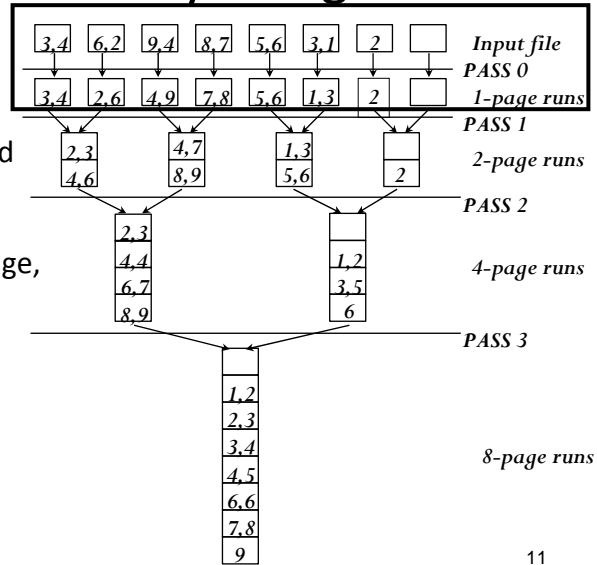
Input file 3,4 6,2 9,4 8,7 5,6 3,1 2

Memory buffer

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A Simple Two-Way Merge Sort

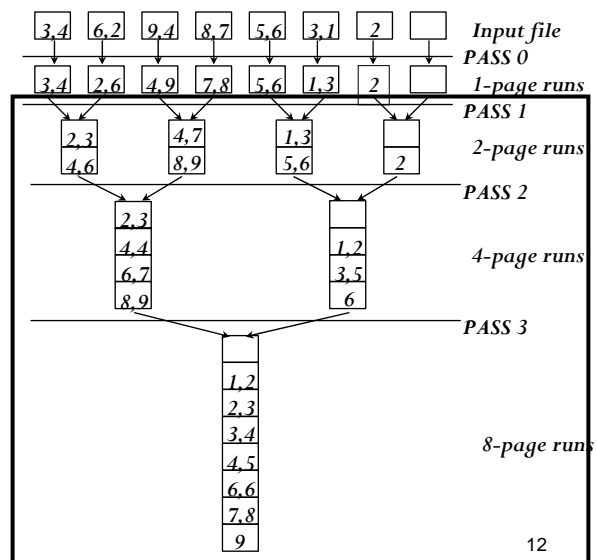
- Basic idea is divide and conquer.
- Sort smaller runs and merge them into bigger runs.
- Pass 0: read each page, sort records in each page, and write the page out to disk. (1 buffer page is used)



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A Simple Two-Way Merge Sort

- Pass 1: read two pages, merge them, and write them out to disk. (3 buffer pages are used)
- Pass 2-3: repeat above step till one sorted 8-page run.
- Each run is defined as a sorted subfile.



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2-Way Merge Sort

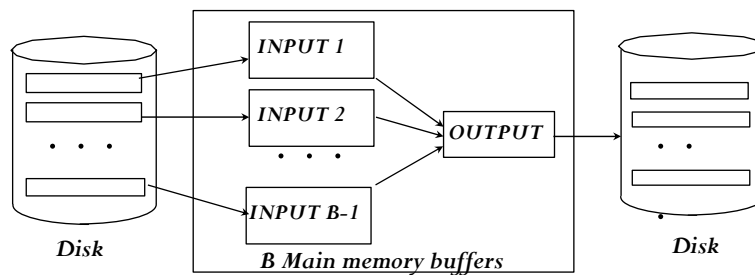
- Say the number of pages in a file is 2^k :
 - Pass 0 produces 2^k sorted runs of one page each
 - Pass 1 produces 2^{k-1} sorted runs of two pages each
 - Pass 2 produces 2^{k-2} sorted runs of four pages each
 - Pass k produces one sorted runs of 2^k pages.
- Each pass requires read + write each page in file: $2*N$
- For a N pages file,
 - the number of passes = $\text{ceiling}(\log_2 N) + 1$
- So total cost (disk I/Os) is
 - $2*N*(\text{ceiling}(\log_2 N) + 1)$

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General External Merge Sort

More than 3 buffer pages. How can we utilize them?

- To sort a file with N pages using B buffer pages:
 - Pass 0: use B buffer pages. Produce N / B sorted runs of B pages each.
 - Pass 1.. k : use $B-1$ buffer pages to merge $B-1$ runs, and use 1 buffer page for output.

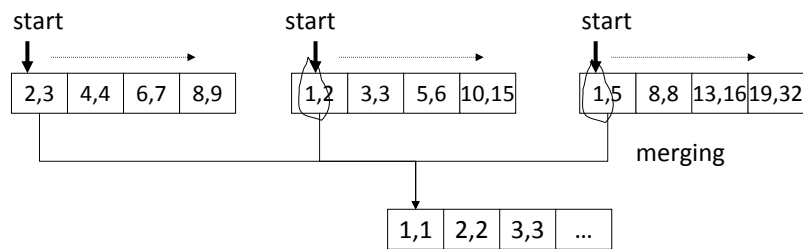


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General External Merge Sort (B=4)

3,4 6,2 9,4 8,7 5,6 3,1 2,10 15,3 16,5 13,8 19,1 32,8

Pass 0: Read four unsorted pages, sort them, and write them out. Produce 3 runs of 4 pages each.



Pass 1: Read three pages, one page from each of 3 runs, merge them, and write them out. Produce 1 run of 12 pages.

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Cost of External Merge Sort

- # of passes: $1 + \text{ceiling}(\log_{B-1} \text{ceiling}(N/B))$
- Disk I/O Cost = $2 * N * (\# \text{ of passes})$
- E.g., with 5 buffer pages, to sort 108 page file:
 - Pass 0: $\text{ceiling}(108/5) = 22$ sorted runs of length 5 pages each (last run is only 3 pages)
 - Pass 1: $\text{ceiling}(22/4) = 6$ sorted runs of length 20 pages each (last run is only 8 pages)
 - Pass 2: 2 sorted runs, of length 80 pages and 28 pages
 - Pass 3: Sorted file of 108 pages
 - # of passes = $1 + \text{ceiling}(\log_4 \text{ceiling}(108/5)) = 4$
 - Disk I/O costs = $2 * 108 * 4 = 864$

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Passes of External Sort

N	$B=3$	$B=5$	$B=9$	$B=17$	$B=129$	$B=257$
100	7	4	3	2	1	1
1,000	10	5	4	3	2	2
10,000	13	7	5	4	2	2
100,000	17	9	6	5	3	3
1,000,000	20	10	7	5	3	3
10,000,000	23	12	8	6	4	3
100,000,000	26	14	9	7	4	4
1,000,000,000	30	15	10	8	5	4

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Further Optimization Possible?

- **Opportunity #1: create bigger length run in pass 0**
 - First sorted run has length B .
 - Is it possible to create bigger length in the first sorted runs?
- **Opportunity #2: consider block I/Os**
 - Block I/O: reading & writing consecutive blocks
 - Can the merge passes use block I/O?
- **Opportunity #3: minimize CPU/disk idle time**
 - How to keep both CPU & disks busy at the same time?

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Opportunity 1: create bigger runs on the 1st pass

3,4 6,2 9,4 8,7 5,6 3,1 2,10 15,3 16,5 13,8 19,1 32,8

Current
Set Buffer
Input Buffer
Output Buffer

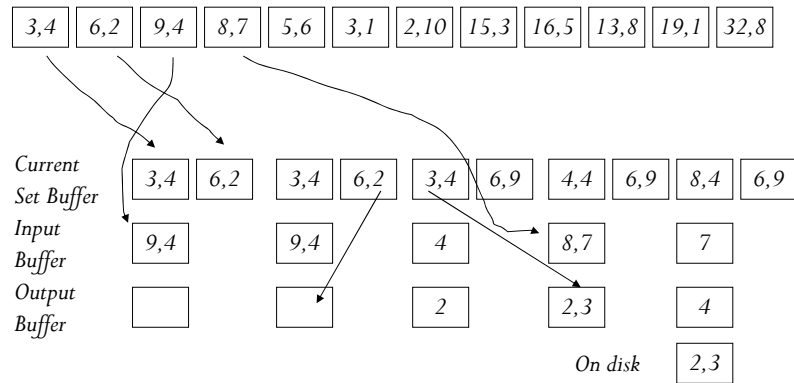
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Replacement Sort (Optimize Merge Sort)

- Pass 0 can output approx. 2B sorted pages on average. How?
- Divide B buffer pages into 3 parts:
 - Current set buffer (B-2): unsorted or unmerged pages.
 - Input buffer (1 page): one unsorted page.
 - Output buffer (1 page): output sorted page.
- Algorithm:
 - Pick the tuple in the current set with the smallest k value > largest value in output buffer.
 - Append k to output buffer.
 - This creates a hole in current set, so move a tuple from input buffer to current set buffer.
 - When the input buffer is empty of tuples, read in a new unsorted page.

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Replacement Sort Example (B=4)



When do you start a new run?

All tuple values in the current set $<$ the last tuple value in output buffer

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Minimizing I/O Cost vs. Number of I/Os

- So far, the cost metric is the number of disk I/Os.
- This is inaccurate for two reasons:
 - (1) Block I/O is a much cheaper (per I/O request) than equal number of individual I/O requests.
 - Block I/O: read/write several consecutive pages at the same time.
 - (2) CPU cost still matters a bit.
 - Keep CPU busy while we wait for disk I/Os.
 - Double Buffering Technique.

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Further Optimization Possible?

- Opportunity #1: create bigger length run in the 1st round
 - First sorted run has length B.
 - Is it possible to create bigger length in the first sorted runs?
- Opportunity #2: consider block I/Os
 - Block I/O: reading & writing consecutive blocks is much faster than separate blocks
 - Can the merge passes use block I/O?
- Opportunity #3: minimize CPU/disk idle time
 - How to keep both CPU & disks busy at the same time?

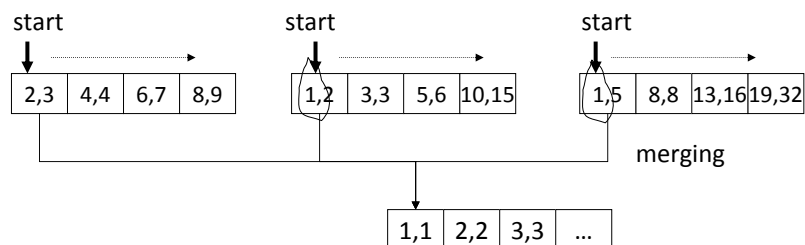
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General External Merge Sort (B=4)

How to change it to Block I/O (block = 2 pages)?

3,4 6,2 9,4 8,7 5,6 3,1 2,10 15,3 16,5 13,8 19,1 32,8

Pass 0: Read four unsorted pages, sort them, and write them out. Produce 3 runs of 4 pages each.



Pass 1: Read three pages, one page from each of 3 runs, merge them, and write them out. Produce 1 run of 12 pages.

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Block I/O

- Block access: read/write b pages as a unit.
- Assume the buffer pool has B pages, and file has N pages.
- Look at cost of external merge-sort (with replacement optimization) using Block I/O:
 - Block I/O has little affect on pass 0.
 - Pass 0 produces initial N' ($= N/2B$) runs of length $2B$ pages.
 - Pass 1..k, we can merge $F = B/b - 1$ runs.
 - The total number of passes (to create one run of N pages) is $1 + \log_F(N')$.

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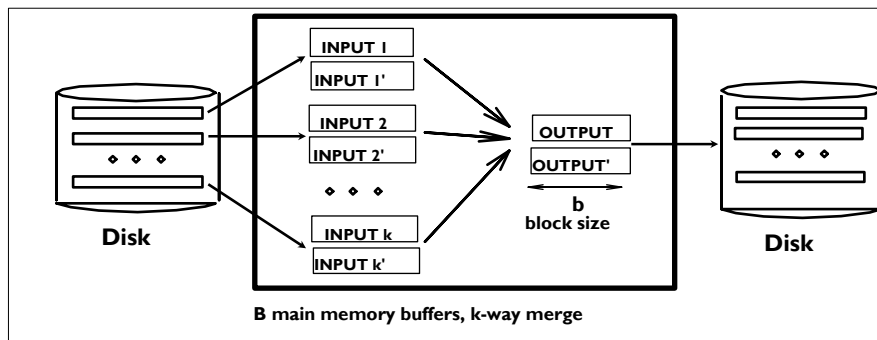
Further Optimization Possible?

- Opportunity #1: create bigger length run in the 1st round
 - First sorted run has length B .
 - Is it possible to create bigger length in the first sorted runs?
- Opportunity #2: consider block I/Os
 - Block I/O: reading & writing consecutive blocks
 - Can the merge passes use block I/O?
- Opportunity #3: minimize CPU/disk idle time
 - While CPU is busy sorting or merging, disk is idle.
 - How to keep both CPU & disks busy at the same time?

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Double Buffering

- Keep CPU busy, minimizes waiting for I/O requests.
 - While the CPU is working on the current run, start to prefetch data for the next run (called shadow blocks).
- Potentially, more passes; in practice, most files still sorted in 2-3 passes.

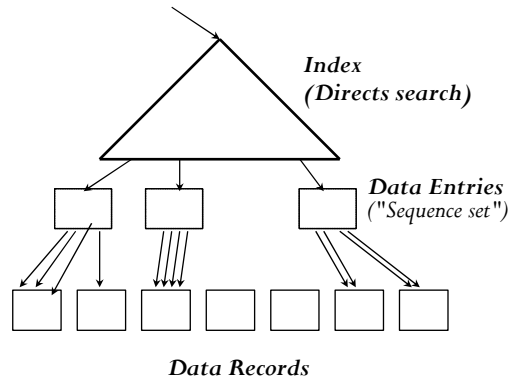


Using B+ Trees for Sorting

- Assumption: Table to be sorted has B+ tree index on sorting column(s).
- Idea: Can retrieve records in order by traversing leaf pages.
- Is this a good idea?
- Cases to consider:
 - B+ tree is clustered
 - B+ tree is not clustered

Clustered B+ Tree Used for Sorting

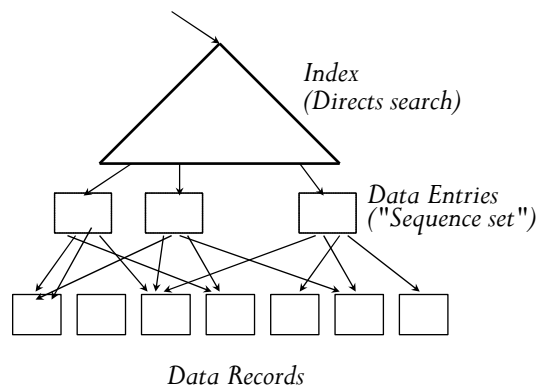
- Cost: root to the left-most leaf, then retrieve all leaf pages (Alternative 1) (Alternative 1)
- If Alternative 2 is used? Additional cost of retrieving data records: each page fetched just once.
- Cost better than external sorting?



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Unclustered B+ Tree Used for Sorting

- Alternative (2) for data entries; each data entry contains rid of a data record. In general, one I/O per data record.



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External Sorting vs. Unclustered Index

N	Sorting	$p=1$	$p=10$	$p=100$
100	200	100	1,000	10,000
1,000	2,000	1,000	10,000	100,000
10,000	40,000	10,000	100,000	1,000,000
100,000	600,000	100,000	1,000,000	10,000,000
1,000,000	8,000,000	1,000,000	10,000,000	100,000,000
10,000,000	80,000,000	10,000,000	100,000,000	1,000,000,000

* p : # of records per page

* $B=1,000$ and block size=32 for sorting

* $p=100$ is the more realistic value. ³¹

We are done with Chapter 13

Chapter 14 (only section 14.4)

Equality Joins With One Join Column



```
SELECT *  
FROM Reserves R1, Sailors S1  
WHERE R1.sid=S1.sid
```

- $R \times S$ is very common, so must be carefully optimized.
- $R \times S$ is large; so, $R \times S$ followed by a selection is inefficient.
- Assume: M pages in R , p_R tuples per page, N pages in S , p_S tuples per page.
 - In our examples, R is Reserves and S is Sailors.
- We will consider more complex join conditions later.

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Two Classes of Algorithms to Implement Join Operation

- Algorithms in class 1 require enumerating all tuples in the cross-product and discard tuples that do not meet the join condition.
 - Simple Nested Loops Join
 - Blocked Nested Loops Join
- Algorithms in class 2 avoid enumerating the cross-product.
 - Index Nested Loops Join
 - Sort-Merge Join
 - Hash Join

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Simple Nested Loops Join

foreach tuple r in R do

 foreach tuple s in S do

 if $r_i == s_j$ then add $\langle r, s \rangle$ to result

- For each tuple in the *outer* relation R, scan the entire *inner* relation S (scan S total of $p_R * M$ times!).
 - Cost: $M + p_R * M * N = 1000 + 100 * 1000 * 500$ I/Os => very huge.
- How can we improve simple nested loops join?

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Page Oriented Nested Loops Join

foreach page of R do

 foreach page of S do

 for all matching tuples r in R-block and

 s in S-page, add $\langle r, s \rangle$ to result

- Cost: $M + M * N = 1000 + 1000 * 500 = 501,000$ => still huge.
- If smaller relation (S) is outer, cost = $500 + 500 * 1000 = 500,500$

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Block Nested Loops Join

foreach block of B-2 pages of R do

 foreach page of S do

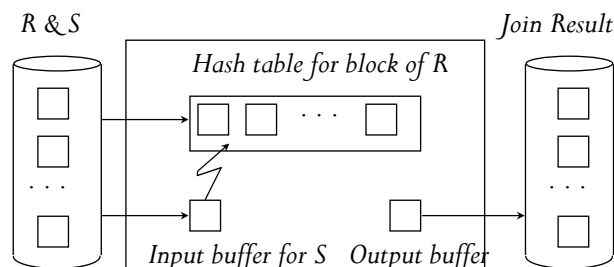
 for all matching tuples r in R-block and
 s in S-page, add $\langle r, s \rangle$ to result

- Use one page as an input buffer for scanning the inner S, one page as the output buffer, and use all remaining pages to hold ``block'' of outer R.
 - For each matching tuple r in R-block, s in S-page, add $\langle r, s \rangle$ to result. Then read next R-block, scan S, etc.

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Block Nested Loops Join: Efficient Matching Pairs

- If B is large, it may be slow to find matching pairs between tuples in S-page and R-block (R-block has B-2 pages).
- The solution is to build a main-memory hash table for R-block.



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Examples of Block Nested Loops

- Cost: Scan of outer + #outer blocks * scan of inner
 - #outer blocks = ceiling (# of pages of outer / blocksize)
- With Reserves (R) as outer, and 102 buffer pages:
 - Cost of scanning R is 1000 I/Os; a total of 10 blocks.
 - Per block of R, scan Sailors (S); 10*500 I/Os.
 - Total cost = 1000 + 10 * 500 = 6000 page I/Os => huge improvement over page-oriented nested loops join.
- With 100-page block of Sailors as outer:
 - Cost of scanning S is 500 I/Os; a total of 5 blocks.
 - Per block of S, we scan Reserves; 5*1000 I/Os.
 - Total cost = 500 + 5*1000 = 5500 page I/Os
- For blocked access (block I/Os are more efficient), it may be best to divide buffers evenly between R and S.

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Index Nested Loops Join

```
foreach tuple r in R do
  foreach tuple s in S where ri == sj do
    add <r, s> to result
```

- If there is an index on the join column of one relation (say S), can make it the inner and exploit the index.
 - Cost: $M + (M * p_r) * \text{cost of finding matching S tuples}$
- For each R tuple, cost of probing S index is about 1.2 for hash index, 2-4 for B+ tree. Cost of finding S tuples (assume Alt. (2) or (3) for data entries) depends on clustering.
 - Clustered index: 1 I/O (typical)
 - Unclustered index: up to 1 I/O per matching S tuple.

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Examples of Index Nested Loops

- Hash-index (Alt. 2) on *sid* of Sailors (as inner):
 - Scan Reserves: 1000 page I/Os, 100*1000 tuples.
 - For each Reserves tuple: 1.2 I/Os to get data entry in index, plus 1 I/O to get (the exactly one) matching Sailors tuple.
 - Total: $1000 + 100,000 * 2.2 = 221,000$ I/Os.
- Hash-index (Alt. 2) on *sid* of Reserves (as inner):
 - Scan Sailors: 500 page I/Os, 80*500 tuples.
 - For each Sailors tuple: 1.2 I/Os to find index page with data entries, plus cost of retrieving matching Reserves tuples.
 - Assume uniform distribution, 2.5 reservations per sailor (100,000 / 40,000). Cost of retrieving them is 1 or 2.5 I/Os depending on whether the index is clustered.
 - Total (Clustered): $500 + 40,000 * 2.2 = 88,500$ I/Os.
- Given choices, put the relation with higher # tuples as inner loop.
- Index Nested Loop performs better than simple nested loop.

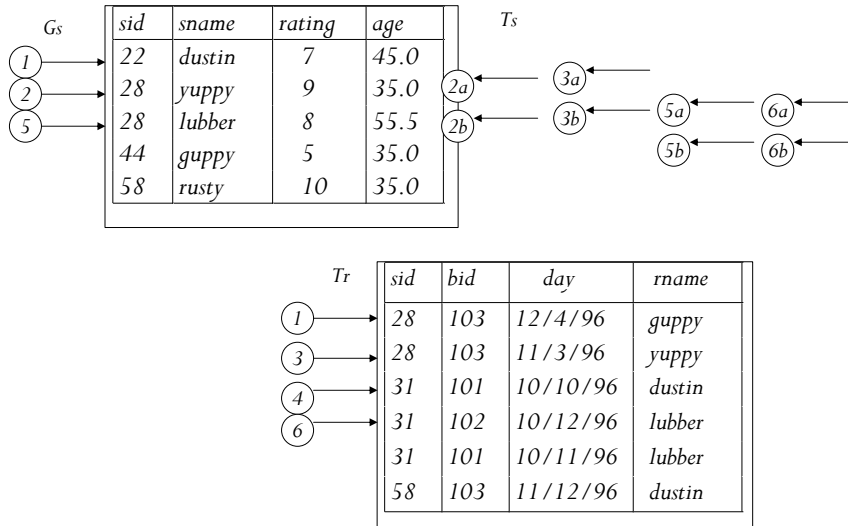
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Sort-Merge Join

- Sort R and S on the join column [merge-sort], then scan them to do a "merge" (on join col.) [scan-merge], and output result tuples.
- Scan-merge:
 - Advance scan of R until current R-tuple \geq current S tuple, then advance scan of S until current S-tuple \geq current R tuple; do this until current R tuple = current S tuple.
 - At this point, all R tuples with same value in R_i (current R group) and all S tuples with same value in S_j (current S group) match; output $\langle r, s \rangle$ for all pairs of such tuples.
 - Then resume scanning R and S.

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Scan-Merge



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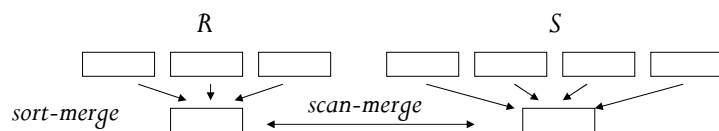
Cost of Merge-Sort Join

- R is scanned once; each S group is scanned once per matching R tuple. (Multiple scans of an S group are likely to find needed pages in buffer.)
- Cost: $2 (M \log M + N \log N) + (M+N)$
 - Assume enough buffer pages to sort both Reserves and Sailors in 2 passes
 - The cost of merge-sorting two relations is $2 M \log M + 2 N \log N$.
 - The cost of scan-merge two sorted relations is $M+N$.
 - Total join cost: $2*2*1000 + 2*2*500 + 1000 + 500 = 7500$ page I/Os.
- Any possible refinement to reduce cost?

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Refinement of Sort-Merge Join

- Combine the merging phase in *sorting* with the scan-merge for the join.
 - Allocate one buffer space for each run (in the merge pass) in R & S.
 - Buffer size $B > \text{squar_root}(L)$, where L is the size of the larger relation. Why?
 - # runs = $2(L/2B) = L/B < B$ [replacement sort]
 - Cost: read+write each relation in Pass 0 + read each relation in (only) merging pass (not counting the writing of result tuples).
 - Cost goes down from 7500 to 4500 I/Os



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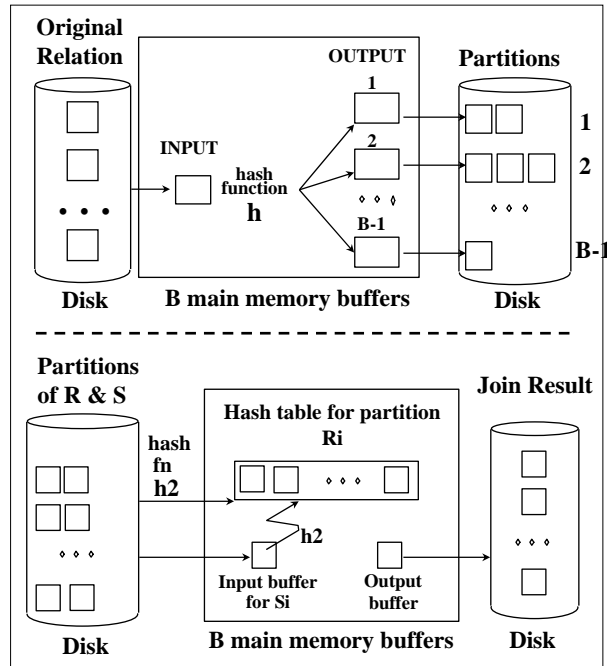
Hash Join

- Hash both relations on the join attribute using the same hash function h .
 - Tuples in R-partition _{i} (bucket) can only match with tuples in S-partition _{i} .
- For $i=1..k$, check for matching pairs in R-partition _{i} and S-partition _{i} .
- For efficient matching pairs, apply hashing to tuples of R-partition using another hash function h_2 .

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Hash-Join

- Partition both relations using hash fn h : R tuples in partition i will only match S tuples in partition i .
- Read in a partition of R , hash it using h_2 ($\leftrightarrow h!$). Scan matching partition of S , search for matches.



Observations on Hash-Join

- #partitions $k < B-1$ (need one buffer page for reading), and $B-2 >$ size of largest partition to be held in memory.
- Assume uniformly sized partitions, and maximize k , we get:
 - $k = B-1$, and $B-2 > M/(B-1)$, i.e., B must be $> \sqrt{M}$
- If we build an in-memory hash table to speed up the matching of tuples, a little more memory is needed.
- If the hash function does not partition uniformly, one or more R partitions may not fit in memory. Can apply hash-join technique recursively to do the join of this R -partition with corresponding S -partition.

Cost of Hash-Join

- In partitioning phase, read+write both relns; $2(M+N)$. In matching phase, read both relns; $M+N$ I/Os.
- In our running example, this is a total of 4500 I/Os.
- Sort-Merge Join vs. Hash Join:
 - Same amount of buffer pages.
 - Same cost of $3(M+N)$ I/Os.
 - Hash Join shown to be highly parallelizable.
 - Sort-Merge less sensitive to data skew; result is sorted.

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Complex Join Conditions

- So far, we have only discussed single equality join condition.
- How about equalities over several attributes? (e.g., $R.sid=S.sid$ AND $R.rname=S.sname$):
 - For Index Nested Loops join, build index on $\langle sid, sname \rangle$ on R (if R is inner); or use existing indexes on sid or sname.
 - For Sort-Merge and Hash Join, sort/partition on combination of the two join columns.

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