Joins Review and Bonus

Nested Loop Joins

Notes

We are again considering "IO aware" algorithms:
 care about disk IO

- Given a relation R, let:
 - T(R) = # of tuples in R
 - P(R) = # of pages in R
- Note also that we omit ceilings in calculations...
 good exercise to put back in!

Given *B+1* pages of memory

```
Compute R⋈S on A:
  for each B-1 pages pr of R:
    for page ps of S:
      for each tuple r in pr:
        for each tuple s in ps:
          if r[A] == s[A]:
            yield (r,s)
```

Cost:

P(R)

1. Load in B-1 pages of R at a time (leaving 1 page each free for S & output)

Given *B+1* pages of memory

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```

Cost:

P(R)+P(R)/B-1P(S)

- Load in B-1 pages of R at a time (leaving 1 page each free for S & output)
- 2. For each (B-1)-page segment of R, load each page of S

This line is called P(R)/B-1 times; the loop iterates over the entire relation S P(R)/B-1 times (ceiling!)

Given *B+1* pages of memory

```
Compute R⋈S on A:
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          if r[A] == s[A]:
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```

Cost:

P(R) + P(R)/B - 1 P(S)

- Load in B-1 pages of R at a time (leaving 1 page each free for S & output)
- 2. For each (B-1)-page segment of R, load each page of S
- 3. Check against the join conditions

BNLJ can also handle non-equality constraints

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```
Compute R⋈S on A:
  for each B-1 pages pr of R:
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            yield (r,s)
```

Cost:

$$P(R) + P(R)/B - 1P(S) + OUT$$

- Load in B-1 pages of R at a time (leaving 1 page each free for S & output)
- 2. For each (B-1)-page segment of R, load each page of S
- 3. Check against the join conditions

4. Write out

BNLJ: Some quick facts.

- We use B+1 buffer pages as:
 - 1 page for S
 - 1 page for output
 - B-1 Pages for R

$$P(R) + P(R)/B - 1 P(S) + OUT$$

- If P(R) <= B-1 then we do one pass over S, and we run in time P(R) + P(S) + OUT.
 - Note: This is optimal for our cost model!
 - Thus, if min {P(R), P(S)} <= B-1 we should always use BNLJ
 - We use this at the end of **hash join.** We define end condition, one of the buckets is smaller than B-1!

1. Sort-Merge Join (SMJ)



Sort Merge Join (SMJ): Basic Procedure

To compute $R \bowtie S$ on A:

1. Sort R, S on A using external merge sort

Note that we are only considering equality join conditions here

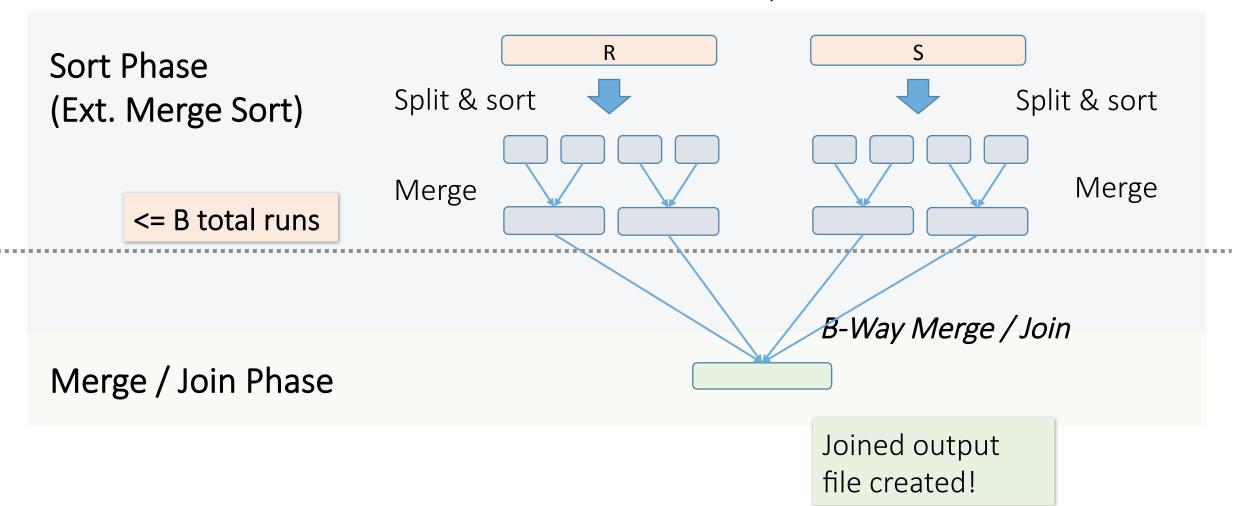
- 2. Scan sorted files and "merge"
- 3. [May need to "backup"- if there are many duplicate join keys]

Note that if R, S are already sorted on A, SMJ will be awesome!

Simple SMJ Optimization

Given *B+1* buffer pages

Unsorted input relations



Simple SMJ Optimization

Given *B+1* buffer pages

- On this last pass, we only do P(R) + P(S) + OUT IOs to complete the join!
- If we can initially split R and S into B total runs each of length approx then
 we only need 3(P(R) + P(S)) + OUT for SMJ!
 - 2 R/W per page to sort runs in memory, 1 R per page to B-way merge / join!
- How much memory for this to happen?
 - $P(R)+P(S)/B \le 2(B+1) \Rightarrow \sim P(R)+P(S) \le 2B \uparrow 2$
 - Thus, $\max\{P(R), P(S)\} \le B12$ is an approximate sufficient condition

If the larger of R,S has \leq B² pages, then SMJ costs 3(P(R)+P(S)) + OUT!



Bonus questions.



- Q1: Fast dog.
 - If max $\{P(R), P(S)\} < B^2$ then SMJ takes 3(P(R) + P(S)) + OUT
 - What is the similar condition to obtain 5(P(R) + P(S)) + OUT?
 - What is the condition for (2k+1)(P(R) + P(S)) + OUT

- Q2: BNLJ V. SMJ
 - Under what conditions will BNLJ outperform SMJ?
 - Size of R, S and # of buffer pages
- Discuss! And We'll put up a google form.

4. Hash Join (HJ)



To compute $R \bowtie S$ on A:

- 1. Partition Phase: Using one (shared) hash function h_B per pass partition R and S into **B** buckets.
 - Each phase creates B more buckets that are a factor of B smaller.
 - Repeatedly partition with a new hash function
 - Stop when all buckets for one relation are smaller than B-1 (Why?)

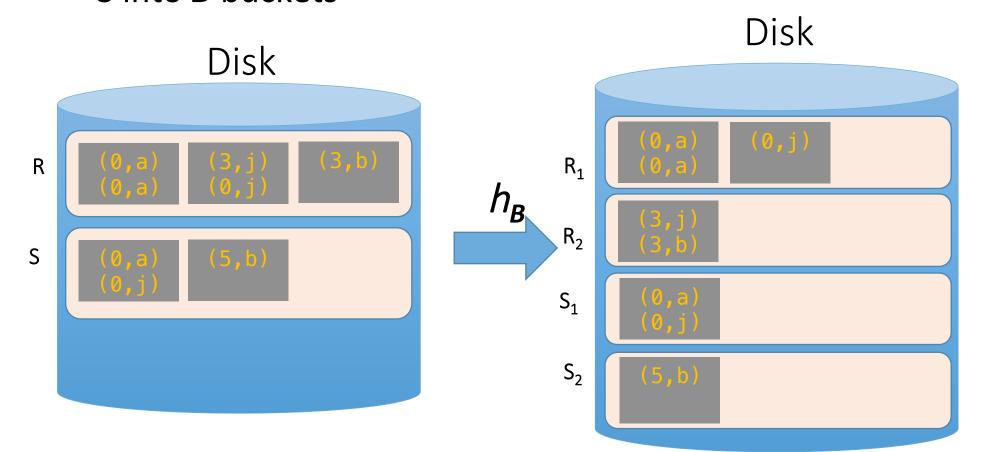
Each pass takes 2(P(R) + P(S))

- 2. Matching Phase: Take pairs of buckets whose tuples have the same values for *h*, and join these
 - Use BNLJ here for each matching pair.

P(R) + P(S) + OUT

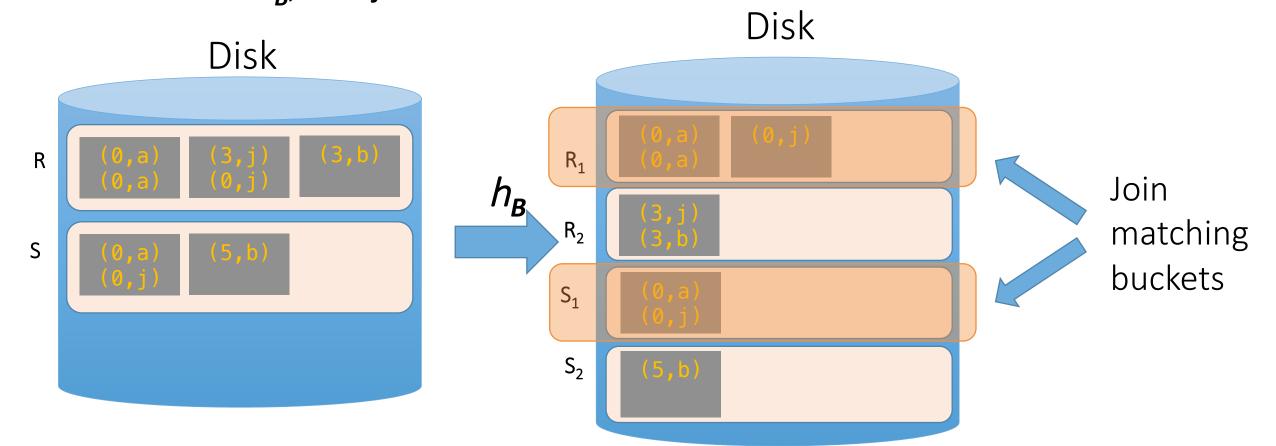
We *decompose* the problem using $h_{\rm B}$, then complete the join

1. Partition Phase: Using one (shared) hash function h_B , partition R and S into **B** buckets

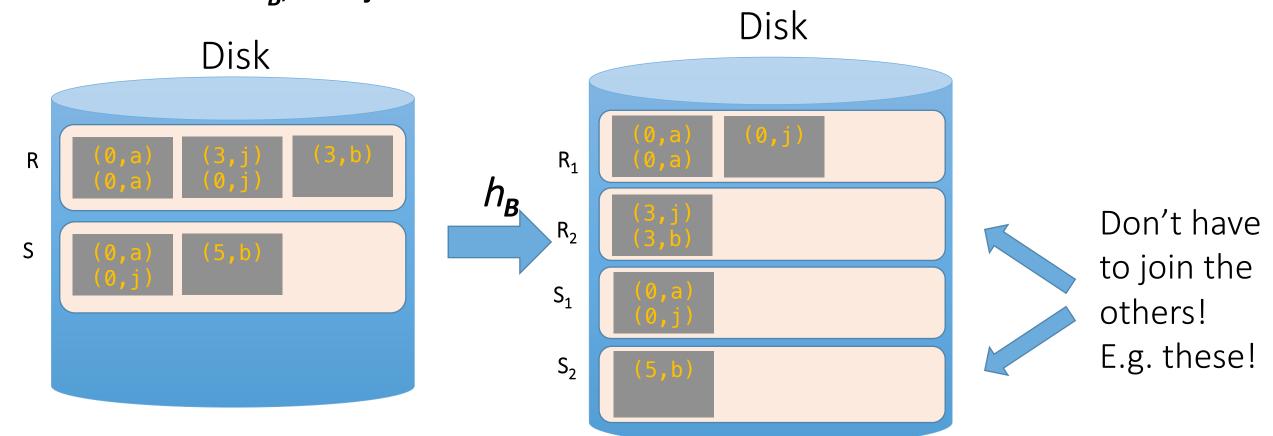


Note our new convention: pages each have two tuples (one per row)

2. Matching Phase: Take pairs of buckets whose tuples have the same values for h_B , and join these



2. Matching Phase: Take pairs of buckets whose tuples have the same values for h_B , and join these





Bonus questions #2



- Q1: Fast little dog.
 - If min $\{P(R), P(S)\} < B^2$ then HJ takes 3(P(R) + P(S)) + OUT
 - What is the similar condition to obtain 5(P(R) + P(S)) + OUT?
 - What is the condition for (2k+1)(P(R) + P(S)) + OUT
- Q2: SMJ V. HJ
 - Under what conditions will HJ outperform SMJ?
 - Under what conditions will SMJ outperform SMJ?
 - Size of R, S and # of buffer pages
- Discuss! And We'll put up a google form.

Sort-Merge v. Hash Join

• Given enough memory, both SMJ and HJ have performance:





- "Enough" memory =
 - SMJ: $B^2 > max\{P(R), P(S)\}$
 - HJ: $B^2 > min\{P(R), P(S)\}$

Hash Join superior if relation sizes differ greatly. Why?

Further Comparisons of Hash and Sort Joins

Hash Joins are highly parallelizable.



Sort-Merge less sensitive to data skew and result is sorted



Summary

- I will ask you to compute costs on the final and PS
 - Walk through the algorithms, you'll be able to compute the costs!
- Memory sizes key in hash versus sort join
 - Hash Join = Little dog (depends on smaller relation)
- Skew is a major factor (more on PS)
- Message: The database can compute IO costs, and these are different than a traditional system.