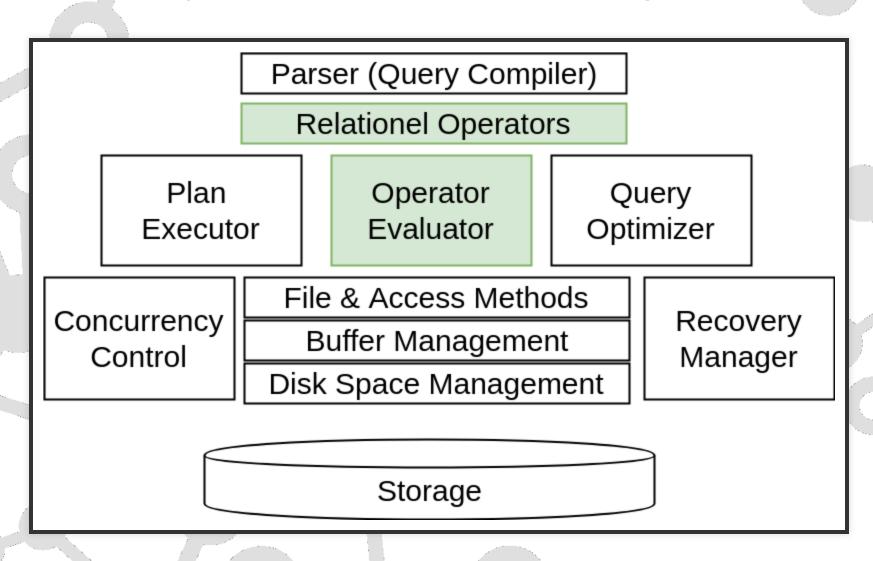
## EVALUATING AN SQL QUERY

### **CHAPTER 12 AND 14**

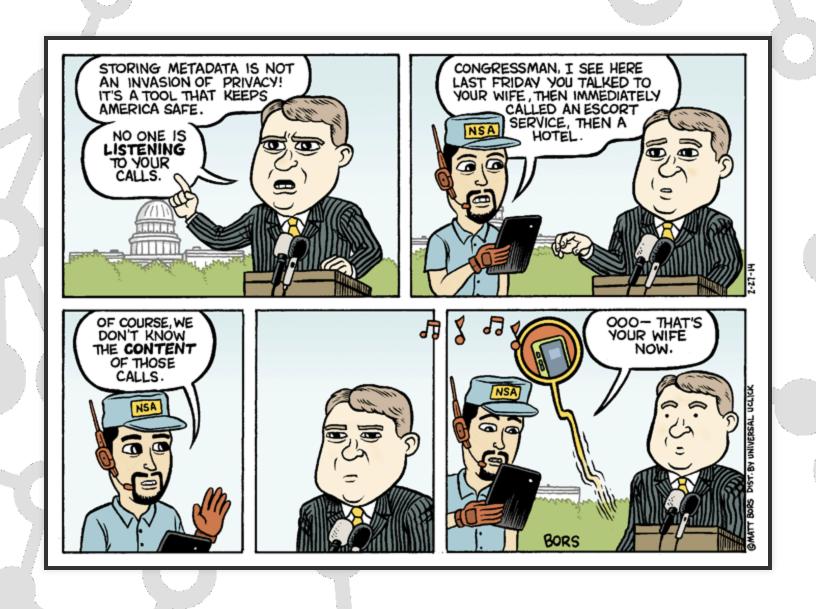
Query Evaluation and Relational Operators



### GOALS

- System Catalog
- Algorithms for algebra operations
- Query evaluation plans and representations
- Intro to query optimization

### THE SYSTEM CATALOG



### INFORMATION IN THE CATALOG

#### For each table:

- Table Name, file name and file structure
- Attribute name and type of all attributes
- index name of each index
- integrity constraints (primary/foreign key constraints)

### INFORMATION IN THE CATALOG

#### For each index:

- Index name and its structure
- Search key attribute

#### For each view:

- View name
- Definition

### INFORMATION IN THE CATALOG

#### Some general statistics:

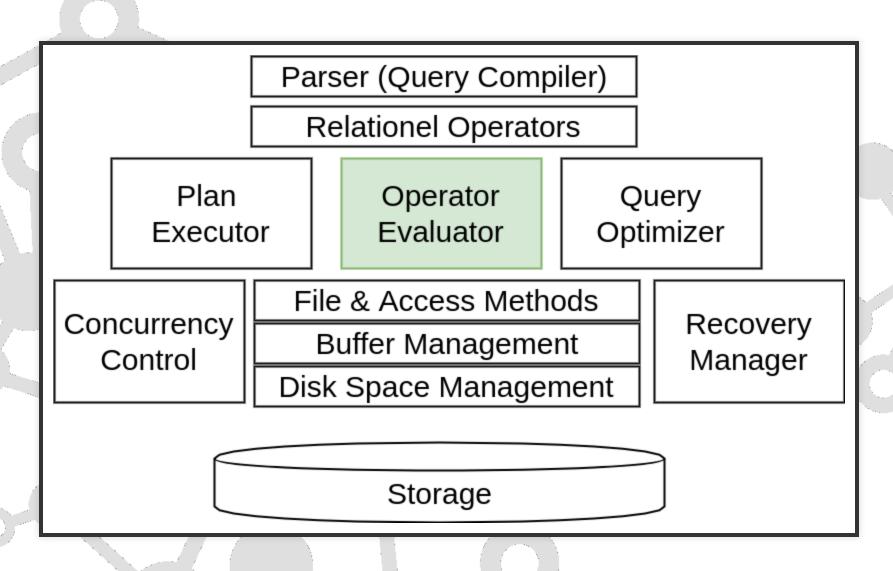
- Cardinality number of tuples in each releation
- Size number of pages for each relation
- Index Cardinality Number of distinct keys in index
- Index size number of pages used for index
- Index Height number of non-leaf levels
- Index range Min and Max key

### HOW CATALOGS ARE STORED

Catalogs are themselves stored as a relation

attr_name	rel_name	type	position
attr_name	Att_Catalog	String	1
rel_name	Att_Catalog	String	2
type	Att_Catalog	String	3
position	Att_Catalog	int	4
sid	Sailor	int	1

# INTRO TO OPERATOR EVALUATION (AND QUERY OPTIMIZATION)



### **OPERATORS**

- Selection ( $\sigma$ ) Select a subset of rows.
- Projection ( $\pi$ ) Remove unwanted columns.
- Join (⋈) Combine two relations.
- Set-difference ( ) Tuples in reln. 1, but not in reln. 2.
- Intersection ( ∩ ) Tuples in both reln. 1 and reln. 2.
- Union (∪) Tuples in rel 1 and/or in reln 2.

### **EXAMPLE QUERY SCHEMA**

#### **Sailors**

Sailors(sid: integer, sname: string, rating: integer, age: real)

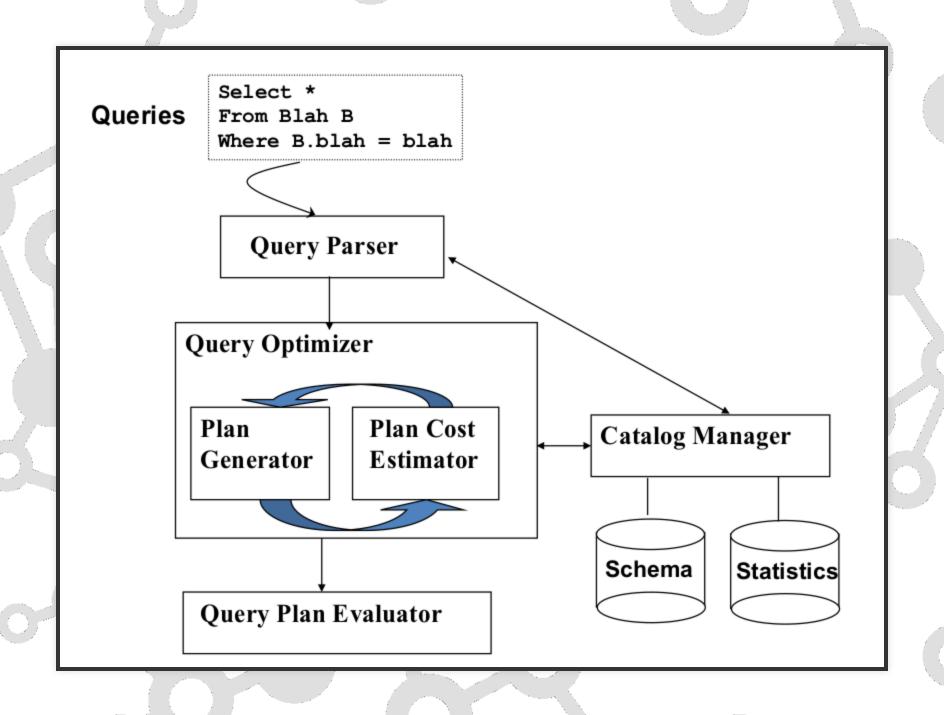
Each tuple is 50 bytes long, 80 tuples per page, 500 pages

#### Reserves

Reserves(sid:integer, bid:integer, day: date, rname: string)

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

### **QUERY SUB-SYSTEM**



### QUERY BLOCKS: UNITS OF OPTIMIZATION

- An SQL query is parsed into a collection of query blocks
  - optimize one block at a time.

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

Outer block Nested block

- Nested blocks are usually treated as calls to a subroutine,
  - called once per outer tuple.

### TRANSLATING SQL TO RELATIONAL ALGEBRA

```
SELECT S.sid, MIN (R.day)
FROM Sailors S, Reserves R, Boats B
WHERE S.sid = R.sid AND R.bid = B.bid AND B.color = "red"
GROUP BY S.sid
HAVING COUNT (*) >= 2
```

For each sailor with at least two reservations for red boats, find the sailor id and the earliest date on which the sailor has a reservation for a red boat.

### TRANSLATING SQL TO RELATIONAL ALGEBRA

```
SELECT S.sid, MIN (R.day)
FROM Sailors S, Reserves R, Boats B
WHERE S.sid = R.sid AND R.bid = B.bid AND B.color = "red"
GROUP BY S.sid
HAVING COUNT (*) >= 2
```

```
\sigma_{S.sid, MIN(R.day)}
(HAVING_{COUNT(*)>2} (
GROUP BY_{S.Sid} (
\sigma_{B.color = "red"} (
Sailors \bowtie Reserves \bowtie Boats))))
S.sid = R.sid R.bid = B.bid
```

### TRANSLATING SQL TO RELATIONAL ALGEBRA



### THE ITERATOR INTERFACE

Relational operators are implemented as iterators:

```
interface iterator {
    void init();
    tuple next();
    void close();
    iterator &inputs[];
    // additional state goes here
}
```

#### Note:

- Edges in the graph are specified by inputs (max 2, usually)
- Any iterator can be input to any other!

```
class Sort extends iterator {
    void init();
    tuple next();
    void close();
    iterator &inputs[1];
    int numberOfRuns;
    DiskBlock runs[];
    RID nextRID[];
}
```

init():

- generate the sorted runs on disk (passes 0 to n-1)
- Allocate runs[] array and fill in with disk pointers.
- Initialize number Of Runs
- Allocate nextRID array and initialize to first RID of each run

next():

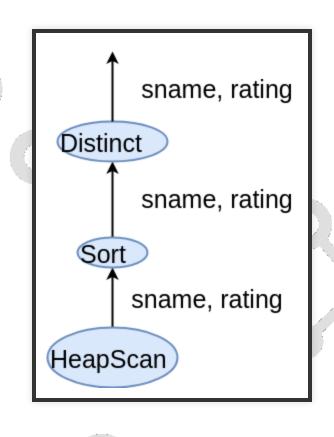
- nextRID array tells us where we're "up to" in each run
- find the next tuple to return based on nextRID array
- advance the corresponding nextRID entry
- return tuple (or EOF "End of Fun" if no tuples remain)

close():

deallocate the runs and nextRID arrays

### QUERY EXECUTION FRAMEWORK

**SELECT DISTINCT** sname, rating **FROM** Sailors



### THREE COMMON TECHNIQUES

- Indexing
- Iteration
- Partitioning

### **ACCESS PATHS**

An access path is a way of retrieving tuples from a table and consists of either

- 1. File scan or
- 2. Index plus matching selection condition

### CONJUNCTIVE NORMAL FORM (CNF)

For a selection condition: attr op value where op is one of <, < =, =, !=, >, >= it is called a **conjunct** 

Conjunctive Normal Form (CNF) is where all the conjuncts are `AND`ed together

### MATCHING A SELECTION CONDITION

An index matches a selection condition if it can be used to retrieve just the tuples that satisfy the condition.

- A hash index matches a CNF selection if there is a term of the form attr
  - = value in the selection for each attribute in the index's search key
- A tree index matches a CNF selection if there is a term of the form attropy value for each attribute in a prefix of the index's search key

An index can match some subset of the conjuncts in a selection condition in CNF. Those are the **primary conjuncts** 

### MATCHING A SELECTION CONDITION

query: day < 8/9/94 AND bid=5 AND sid=3

- 1. Can B+tree index on day be used?
- 2. How about a B+tree on <rname,day>?
- 3. How about a B+tree on <day, rname>?
- 4. How about a hash index on <bid, sid>?
- 5. How about a Hash index on <day, rname>?

### SELECTIVITY OF ACCESS PATHS

The **selectivity** of an access path is the number of pages retrieved (index pages plus data pages) if we use this access path to retrieve all desired tuples.

If a table contains an index that matches a given selection there are at least 2 access paths:

- 1. The index
- 2. Scan the whole file

Sometimes we can scan the index only

### MOST SELECTIVE ACCESS PATHS

The most selective access path is the one that retrieves the fewest pages.

Using this will minimize the cost of data retrieval

### **REDUCTION FACTOR**

The fraction of tuples that satisfy a given conjunct is called reduction factor

When there are several primary conjuncts, the fraction that satisfy them all can be estimated by multiplying their reduction factors

### ALGORITHM FOR RELATIONAL OPERATIONS

We considered Sort last week - lets dive into all the others

### **SELECTION**

The selection operation is a simple retrieval of tuple from a table and implementation follows the lines of access path.

If we have  $\sigma_{R.attr OP \ value}(R)$ 

```
SELECT *
FROM Reserves R
WHERE R.rname < 'C%'
```

### SELECTION - PERFORMANCE

If no appropriate index exists: Must scan the whole relation

cost = |R|

For "reserves" = 1000 I/Os.

### **SELECTION - PERFORMANCE**

With index on selection attribute:

- 1. Use index to find qualifying data entries
- 2. Retrieve corresponding data records

Total cost = cost of step 1 + cost of step 2

For "Reserves", if selectivity = 10% (100 pages, 10000 tuples):

- If clustered index, cost is a little over 100 I/Os
- If unclustered, could be up to 10000 I/Os! ... unless ...

### REFINEMENT FOR UNCLUSTERED INDEXES

- 1. Find qualifying data entries.
- 2. Sort the rids of the data records to be retrieved.
- 3. Fetch rids in order.

Each data page is looked at just once

1. though # of such pages likely to be higher than with clustering

### SELECTION

Best performance depends on:

- What indexes available
- Expected size of result

Size of result: (size of R) \* selectivity

Rule of thumb: It is probably cheaper to simply scan the entire table instead of using an unclustered index if over 5% of the tuples are to be retrieved

- 1. Find the most selective access path
- 2. Retrieve tuples using it
- 3. Apply any remaining terms that don't match the index

query: day < 8/9/94 AND bid=5 AND sid=3

Some options:

- B+tree index on day; check bid=5 and sid=3 afterward.
- hash index on <bid, sid>; check day < 8/9/94 afterward.

use 2 or more matching indexes.

- 1. From each index, get set of rids
- 2. Compute intersection of rid sets
- 3. Retrieve records for rids in intersection
- 4. Apply any remaining terms

query: day < 8/9/94 AND bid=5 AND sid=3

Suppose we have an index on day, and another index on sid.

- Get rids of records satisfying day<8/9/94.</li>
- Also get rids of records satisfying sid=3.
- Find intersection, then retrieve records
- Then check bid=5.

# **PROJECTION**

Orop certain fields of the input - quite easy

# **PROJECTION**

If DISTINCT is not in the SELECT clause of the SQL, it is trivial to drop the columns of the output that is not needed.

#### The issue is removing duplicates

**SELECT DISTINCT** R.sid, R.bid **FROM** Reserves R

Typically, we use a partitioning technique.

# PROJECTION USING SORTING

- 1. Scan R, extract only the needed attributes
- 2. Sort the resulting set
- 3. Remove adjacent duplicates

#### Cost:

Reserves with size ratio 0.25 = 250 pages and there are 20 buffer pages

With 20 buffer pages can sort in 2 passes, so:

Total cost = 1000 + 250 + 2 \* 2 \* 250 + 250 = 2500 I/Os

# PROJECTION - IMPROVED

Modify the external sort algorithm:

- Modify Pass 0 to eliminate unwanted fields.
- Modify Passes 1+ to eliminate duplicates.

#### Cost:

Reserves with size ratio 0.25 = 250 pages.

With 20 buffer pages can sort in 2 passes

- 1. Read 1000 pages
- 2. Write 250 (in runs of 40 pages each)
- 3. Read and merge runs

Total cost = 1000 + 250 + 250 = 1500 I/Os.

# OTHER PROJECTION TRICKS

If a B+Tree index search key prefix has all wanted attrs:

- Scan index in order
- Compare adjacent tuples on the fly (no sorting required!)

If an index search key contains all wanted attrs:

- Do index-only scan
- Apply projection techniques to data entries (much smaller!)

### PROJECTION BASED ON HASHING

#### Idea:

- Many of the things we use sort for don't exploit the order of the sorted data
  - removing duplicates in DISTINCT
  - finding matches in JOIN

Often good enough to match all tuples with equal values

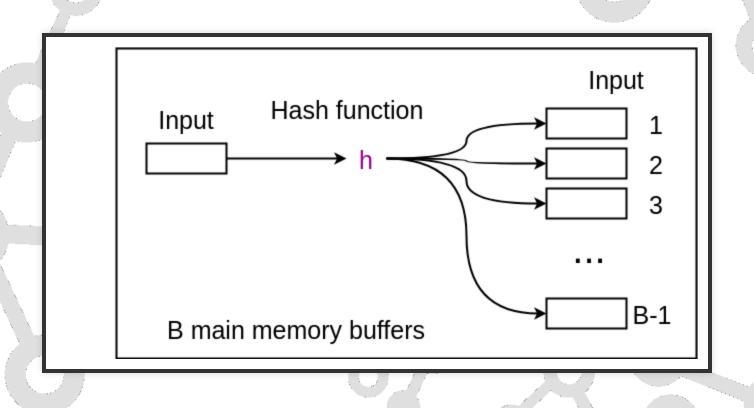
Hashing does this!

And may be cheaper than sorting! (Hmmm...!)

# PROJECTION BASED ON HASHING

**Duplicate Elimination** 

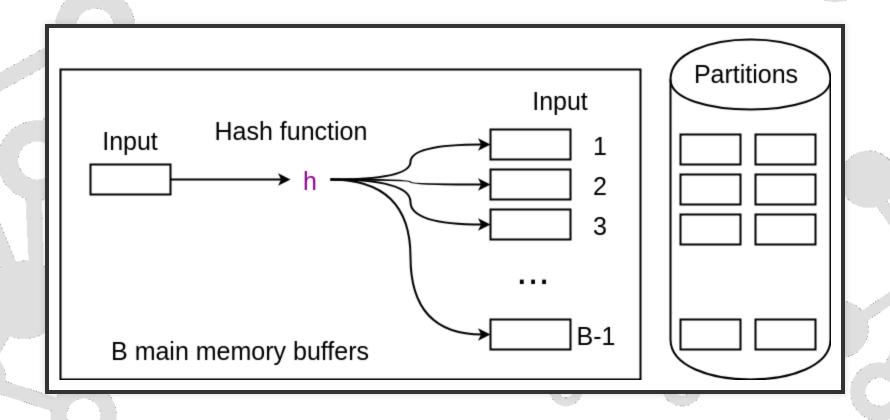
- 1. Apply hash function
- 2. Look for duplicates in the corresponding bucket
- 3. If the input buffer is empty, then read in a page and goto 1.



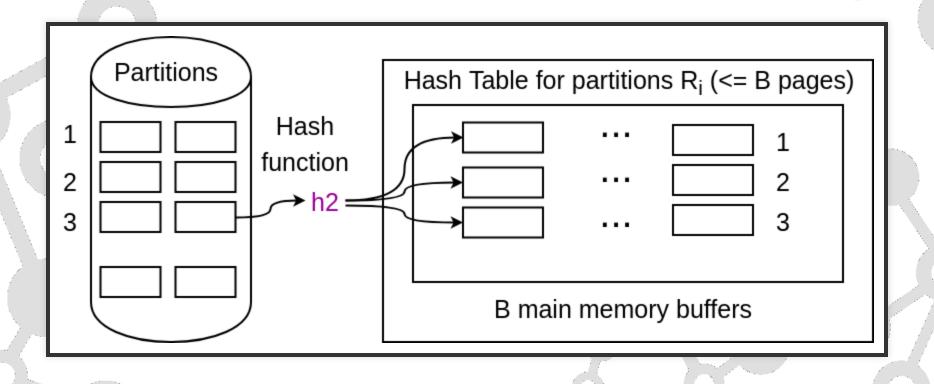
#### Two phases:

- 1. Partition: use a hash function h to split tuples into partitions on disk.
  - Key property: all matches live in the same partition.
- 2. **ReHash:** for each partition on disk, build a main-memory hash table using a hash function h2

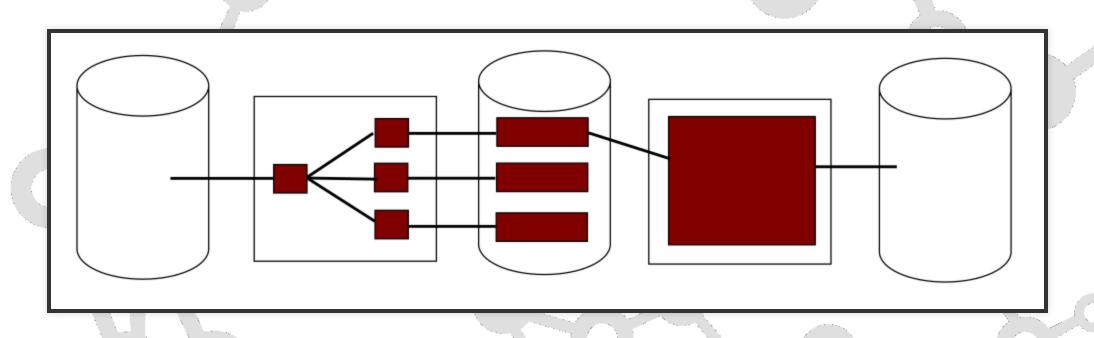
#### **Partition**



#### Rehash

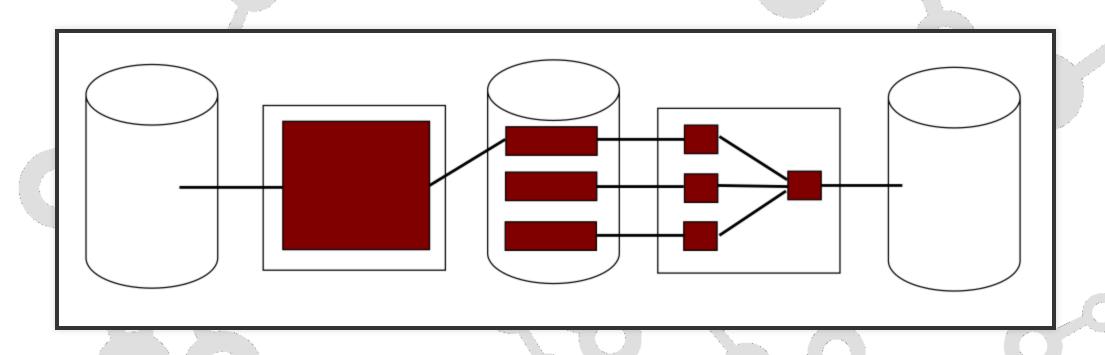


# **COST OF EXTERNAL HASHING**



cost = 3\*|R| 10's

# HOW DOES THIS COMPARE WITH EXTERNAL SORTING?



cost = 3\*|R|IO's

# HOW ABOUT THE MEMORY REQUIREMENTS?

# MEMORY REQUIREMENT FOR EXTERNAL HASHING

How big of a table can we hash in two passes?

- B-1 "partitions" result from Phase 0
- Each should be no more than B pages in size
- Answer: B(B-1).

Said differently:

- We can hash a table of size N pages in about  $\sqrt{N}$  space
  - Note: assumes hash function distributes records evenly!

# MEMORY REQUIREMENT FOR EXTERNAL SORTING

How big of a table can we **sort** in two passes?

- Each "sorted run" after Phase 0 is of size B
- Can merge up to B-1 sorted runs in Phase 1
- Answer: B(B-1).

Said differently:

• We can sort a table of size N pages in about  $\sqrt{N}$  space

# SORTING VS HASHING FOR PROJECTIONS

So which is better ??

Based on our simple analysis:

- Same memory requirement for 2 passes ( $\sqrt{N}$ )
- Same IO cost

Digging deeper ...

### SORTING VS HASHING FOR PROJECTIONS

#### **Sorting pros:**

- Great if input already sorted (or almost sorted)
- Great if need output to be sorted anyway
- Not sensitive to "data skew" or "bad" hash functions

### Hashing pros:

- Highly parallelizable
- Can exploit extra memory to reduce # IOs (stay tuned...)

# JOIN

Joins are very common.

```
SELECT *
```

FROM Reserves R1, Sailors S1 WHERE R1.sid = S1.sid

 $R \times S$  is large; so,  $R \times S$  followed by a selection is inefficient.

Many approaches to reduce join cost - lets have a look.

# SIMPLE NESTED LOOPS JOIN

Tuple at a time nested loops evaluation:  $R \bowtie S$ :

```
foreach tuple r in R do
    foreach tuple s in S do
        if r.sid == s.sid then
        add <r, s> to result
```

Cost =  $(p_R |R|)|S| + |R| = 100 * 1000 * 500 + 1000 |Os$ 

At 10ms/IO ⇒ Total time ~6 days

- What if smaller relation (S) was "outer"?
- What is the cost if one relation can fit entirely in memory?

# PAGE-ORIENTED NESTED LOOPS JOIN

Lets do a page at a time!

foreach page bR in R do
 foreach page bS in S do
 foreach tuple r in bR do
 foreach tuple s in bS do
 if r == s then
 add <r, s> to result

### PAGE-ORIENTED NESTED LOOPS JOIN

Cost =  $|R|^*|S| + |R| = 1000 * 500 + 1000$ 

If smaller relation (S) is outer, cost = 500 \* 1000 + 500

Much better than naive per-tuple approach!

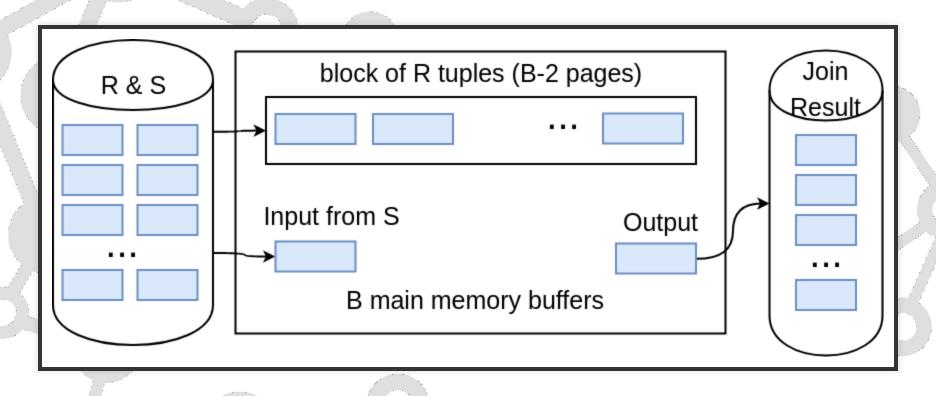
At 10ms/IO ⇒ total time ~ 1.4 hour

The trick is to reduce the # complete reads of the inner table

### **BLOCK NESTED LOOPS JOIN**

Page-oriented NL doesn't exploit extra buffers :(

Idea to use memory efficiently:



Cost: Scan outer + (#outer blocks \* scan inner)

#outer blocks = [ # pages of outer / blocksize ]

# **BLOCK NESTED LOOPS JOIN EXAMPLE**

Say we have B = 100+2 memory buffers

Join cost = |outer| + (#outer blocks \* |inner|)

#outer blocks = |outer| / 100

# BLOCK NESTED LOOPS JOIN EXAMPLE

With R as outer (|R| = 1000):

- Scanning R costs 1000 IO's (done in 10 blocks)
- Per block of R, we scan S; costs 10 \* 500 I/Os
- Total = 1000 + 10 \* 500.
- At 10ms/IO, total time: ~ 1 minute

# BLOCK NESTED LOOPS JOIN EXAMPLE

With S as outer (|S| = 500):

- Scanning S costs 500 IO's (done in 5 blocks)
- Per block of S, we scan R; costs 5\*1000 IO's
- Total = 500 + 5\*1000.
- At 10ms/IO, total time: ~ 55 seconds

# **BLOCKED ACCESS AND DOUBLE-BUFFERING**

Taking blocked acces into account, best approach is to split buffer pool evenly between outer and inner.

- Will result in more page fetching
- Signifcant reduce seeking for pages
- © Could also use double buffering (like in external sort)

# INDEX NESTED LOOPS JOIN

#### $R\bowtie S$ :

foreach tuple r in R do
 foreach tuple s in S where r.att1 == s.att2 do
 add <r, s> to result

- Outer loop use file scan
- Inner loop uses index to find matching tuples

Cost =  $|R| + (|R|*p_R)*$  cost to find matching S tuples

### COST OF INDEX NESTED LOOPS JOIN

- If index uses Alt. 1, cost = cost to traverse tree from root to leaf.
- For Alt. 2 or 3:
  - 1. Cost to lookup RID(s); typically 2-4 IO's for B+Tree.
  - 2. Cost to retrieve records from RID(s); depends on clustering.
    - Clustered index: 1 I/O per page of matching S tuples.
    - Unclustered: up to 1 I/O per matching S tuple.

# **SORT-MERGE JOIN**

- 1. Sort R on join attr(s)
- 2. Sort S on join attr(s)
- 3. Scan sorted-R and sorted-S in tandem, to find matches

Cost: Sort R + Sort S + (|R| + |S|)

In the worst case, last term could be |R|\*|S| (very unlikely!)

Question: what is worst case?

# **COST OF SORT-MERGE JOIN**

Suppose B = 35 buffer pages:

- Both R and S can be sorted in 2 passes
- Total join cost = 4\*1000 + 4\*500 + (1000 + 500) = 7500

Suppose B = 300 buffer pages:

- Again, both R and S sorted in 2 passes
- Total join cost = 7500

# SORT-MERGE JOIN REFINEMENT

Do the join during the final merging pass of sort!

If have enough memory, can do:

- 1. Read R and write out sorted runs
- 2. Read S and write out sorted runs
- 3. Merge R-runs and S-runs, and find R ⋈ S matches

$$Cost = 3*|R| + 3*|S|$$

#### **SORT-MERGE JOIN**

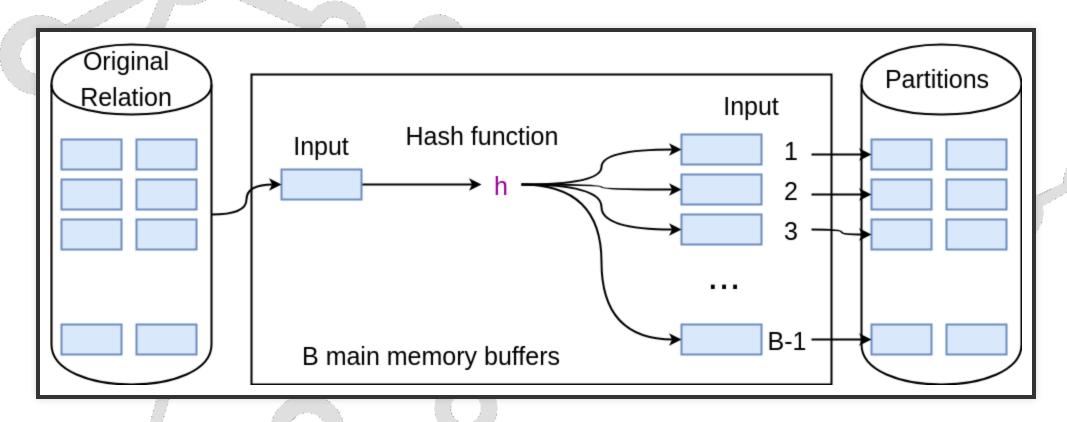
Sort-merge join an especially good choice if:

- one or both inputs are already sorted on join attribute(s)
- output is required to be sorted on join attribute(s)

#### HASH JOIN

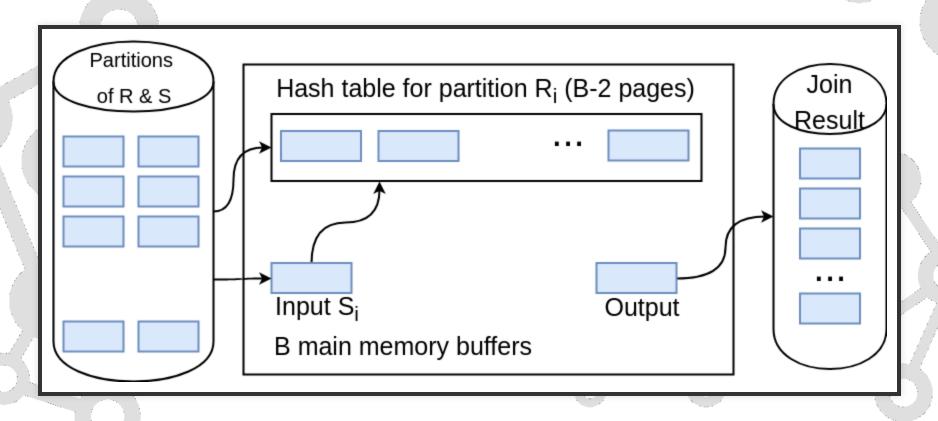
Use idea for large dataset: Partitioning and Matching

Partitioning phase: read + write both relations  $\Rightarrow 2(|R|+|S|)$  I/Os



#### HASH JOIN

Matching phase: read both relations  $\Rightarrow$  |R|+|S| I/Os



Total cost of 2-pass hash join = 3(|R|+|S|)

#### MEMORY REQUIREMENTS AND OVERFLOW HANDLING

- Q: how much memory needed for 2-pass hash join?
- what is cost of 2-pass sort merge join?
- how much memory needed for 2-pass sort merge join?

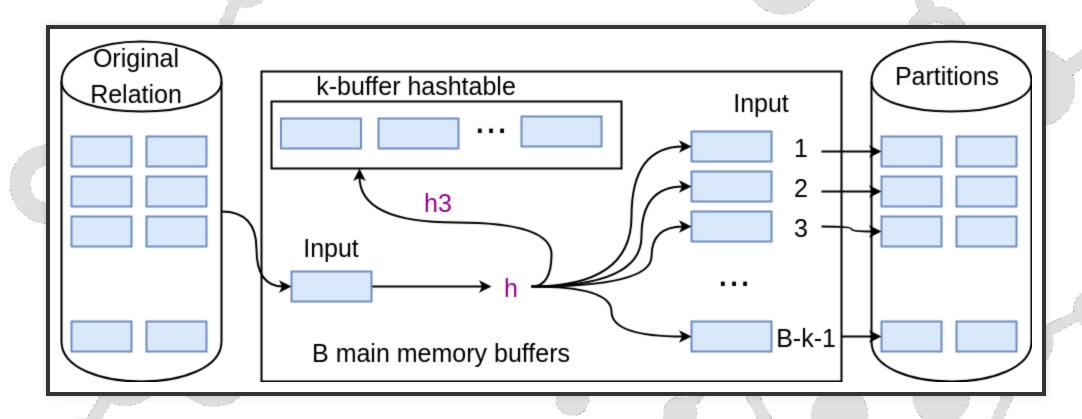
#### UTILIZING EXTRA MEMORY: HYBRID HASH JOIN

Have B memory buffers

Want to hash relation of size N

If  $B < N < B^2$ , we will have unused memory ...

#### HYBRID HASH JOIN



k has to be greater than the size of a partition:  $k \ge \frac{N}{B-k-1}$ 

#### FURTHER DETAILS OF HYBRID HASHING

If we have even more memory, then we can put more than one bucket in the memory

In general, if we can find  $(k \le B)$ , such that  $k \ge \frac{mN}{B-k-1}$ , then we can put m buckets in the memory, while the remaining B-k-1-m buckets are stored on the disk.

#### HASH JOIN VS SORT-MERGE JOIN

#### **Sorting pros:**

- Good if input already sorted, or need output sorted
- Not sensitive to data skew or bad hash functions

#### Hashing pros:

- Often cheaper due to hybrid hashing
- For join: # passes depends on size of smaller relation
- Highly parallelizable

## OTHER OPERATIONS

#### **GROUPING AND AGGREGATION**

© Can you modify the projection algorithm to perform Grouping and Aggregation?

SELECT R.sid, Count(\*)
FROM Reserves R
GROUP BY R.sid

#### GROUPING AND AGGREGATION

Here is the algorithm - what should be modified?

- 1. Scan R, extract only the needed attributes
- 2. Sort the resulting set
- 3. Remove adjacent duplicates

Modify the external sort algorithm:

- Modify Pass 0 to eliminate unwanted fields.
- Modify Passes 1+ to eliminate duplicates.

#### **GROUPING AND AGGREGATION**

- 1. Generate sorted (on the grouping attributes) runs (pass 0 to n-1)
- 2. Get the tuple with the least key v value and start a new group
  - 1. Prepare to compute the aggregates for the group
  - 2. Get all the tuples with key v and update the aggregates of the current group
    - min, max
    - count
    - sum
    - avg
  - 3. If a buffer becomes empty, read from the disk
- 3. If there are more tuples goto 2.

#### SET OPERATIONS

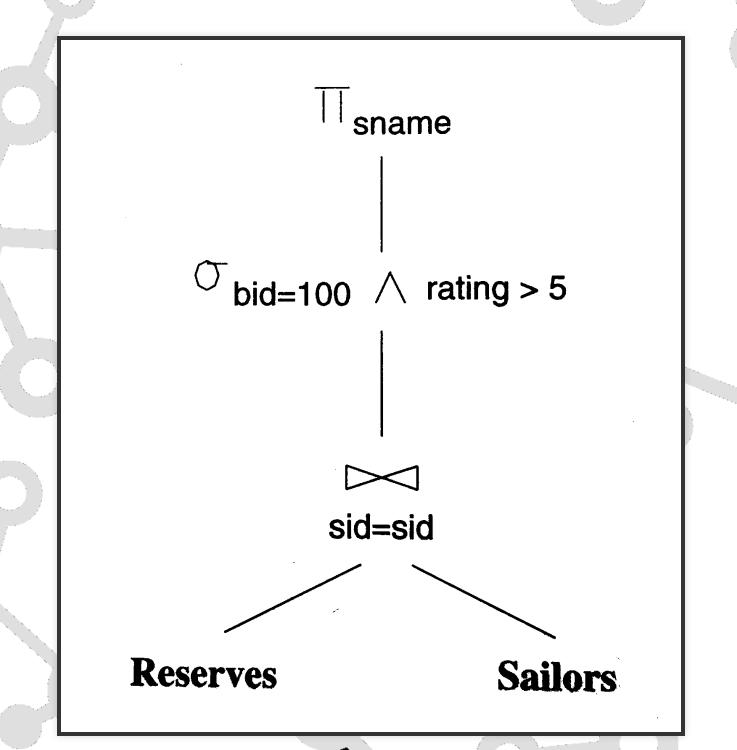
Question: How can we perform union, difference and intersection?

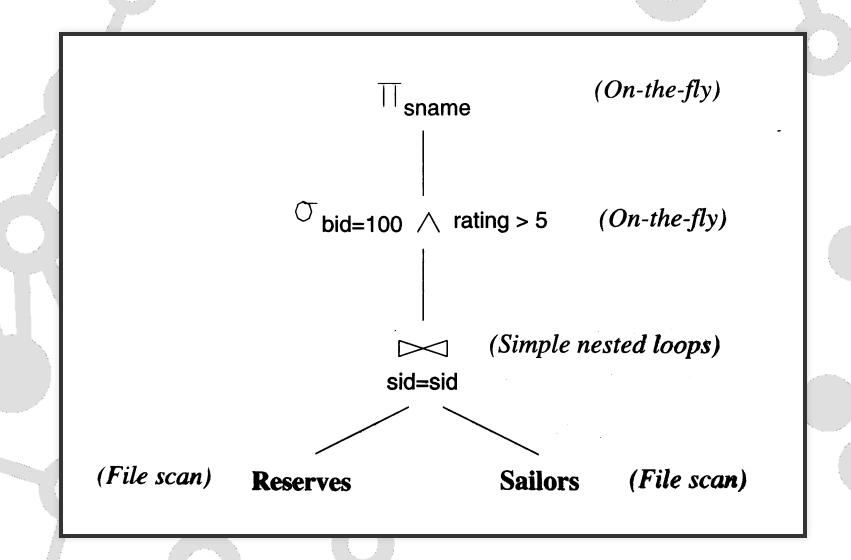
Intro to Query Optimization

SELECT S.sname
FROM Reserves R, Sailors S
WHERE R.sid = S.sid
AND R.bid = 100 AND S.rating > 5

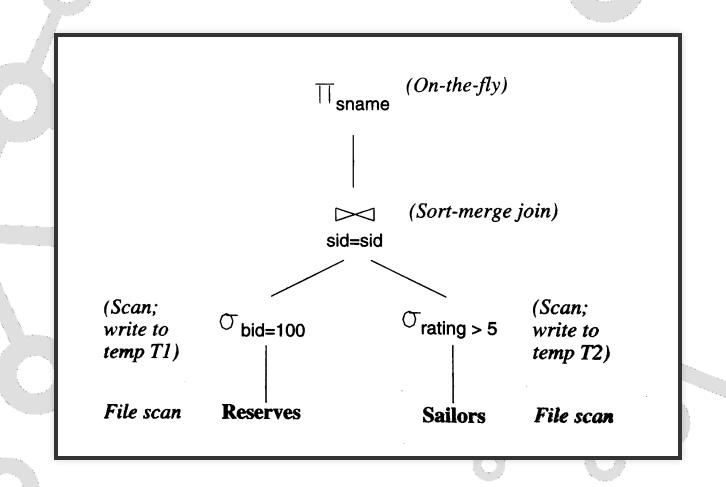
As relational algebra

 $\pi_{sname}(\sigma_{bid \land rating > 5}(Reserves \bowtie_{sid=sid} Sailors))$ 

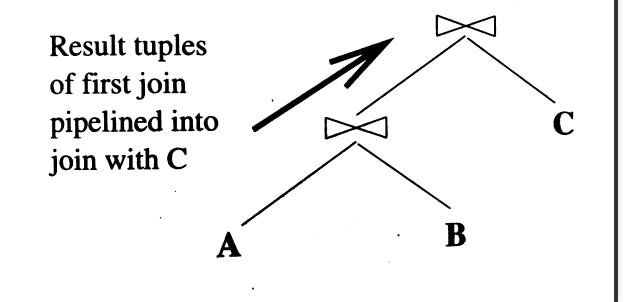




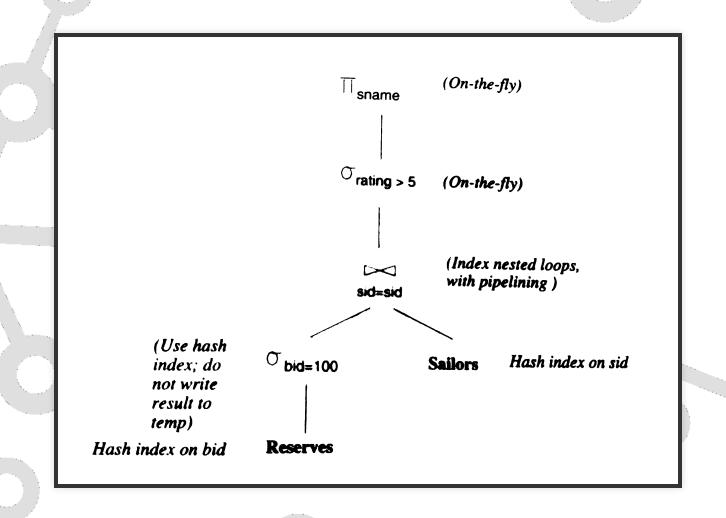
### **PUSHING SELECTIONS**



#### PIPELINING



#### **USING INDEXES**



#### **OPTIMIZATION**

Much more next time!

#### **SUMMARY**

- A virtue of relational DBMSs: queries are composed of a few basic operators
- The implementation of these operators can be carefully tuned
  - Sorting is an important technique for evaluating many operators
  - Index could be exploited in many cases too (be careful with unclustered index)
- Many alternative implementation techniques for each operator
  - No universally superior technique for most operators.
- Must consider available alternatives
  - Called Query optimization we will study this topic next!

# QUESTIONS?