

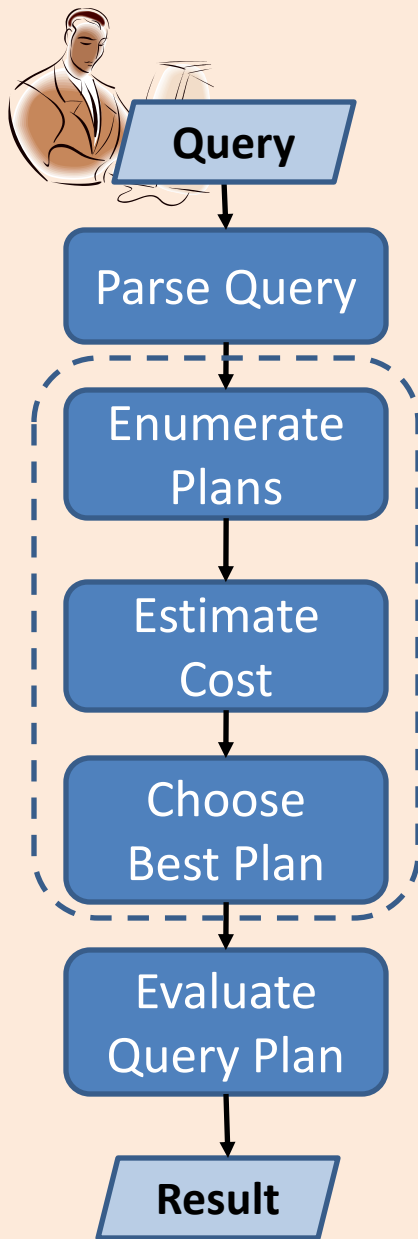
ICS 421 Spring 2010

Relational Query Optimization

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Query Optimization

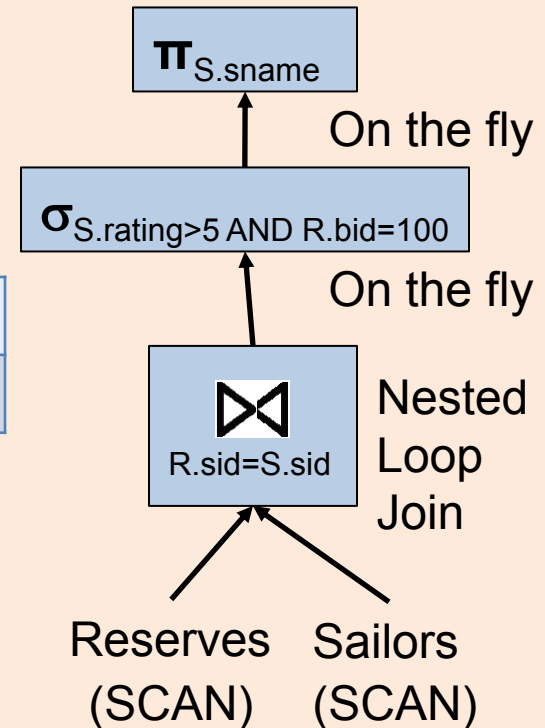
- Two main issues:
 - For a given query, **what plans are considered?**
 - How is the **cost of a plan estimated?**
- **Ideally:** Want to find best plan.
Practically: Avoid worst plans!
- System R Optimizer:
 - Most widely used currently; works well for < 10 joins.
 - **Statistics**, maintained in system catalogs, used to estimate cost of operations and result sizes.
 - Only the space of **left-deep plans** is considered.
 - Cartesian products avoided.

Example

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

Reserves	40 bytes/tuple	100 tuples/page	1000 pages
Sailors	50 bytes/tuple	80 tuples/page	500 pages

- Nested Loop Join cost 1K+
100K*500
- On the fly selection and project does not incur any disk access.
- Total disk access = 500001K
(worst case)



What about complex queries ?

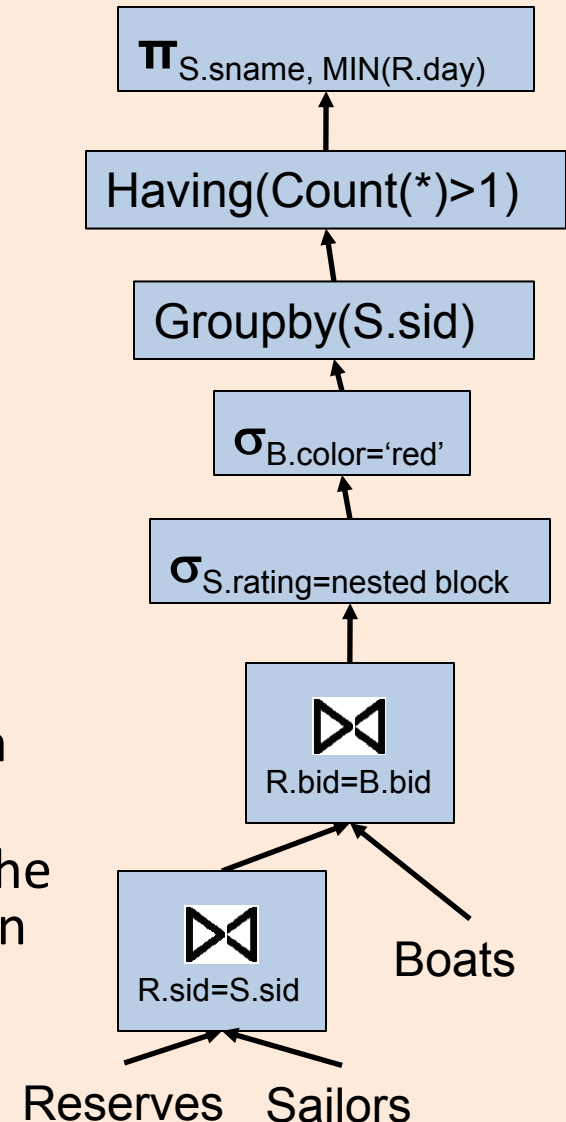
```

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' AND S.rating =
    (SELECT MAX(S2.rating)
     FROM Sailors S2)
GROUP BY S.sid
HAVING COUNT(*) > 1
    
```

Outer Block

Nested Block

- For each block, the plans considered are:
 - All available access methods, for each reln in FROM clause.
 - All left-deep join trees (i.e., all ways to join the relations one-at-a-time, with the inner reln in the FROM clause, considering all reln permutations and join methods.)



RA Equivalences

- **Selections**

- Cascade: $\sigma_{c_1 \wedge \dots \wedge c_n}(R) \equiv \sigma_{c_1}(\dots \sigma_{c_n}(R)\dots)$
- Commute: $\sigma_{c_1}(\sigma_{c_n}(R)) \equiv \sigma_{c_n}(\sigma_{c_1}(R))$

- **Projections**

- Cascade: $\pi_{c_1}(R) \equiv \pi_{c_1}(\dots \pi_{c_n}(R)\dots)$, c_1 subset c_i , $i > 1$

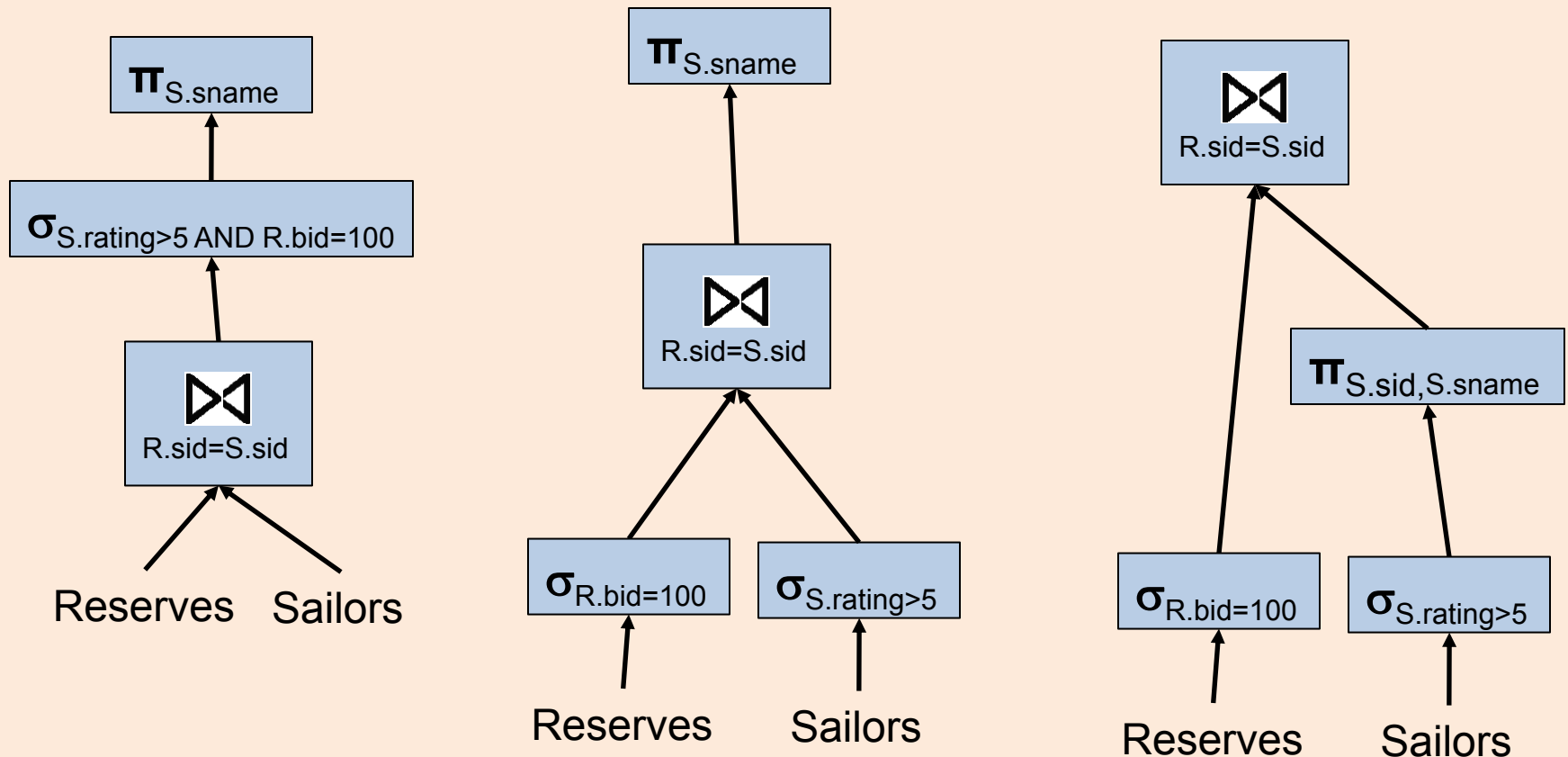
- **Joins**

- Associativity: $R \text{ join } (S \text{ join } T) \equiv (R \text{ join } S) \text{ join } T$
- Commutative: $R \text{ join } S \equiv S \text{ join } R$
- Definition: $R \text{ join } S \equiv \sigma_{R.col=S.col}(R \times S)$

More equivalences

- Commutability between projection & selection
 - $\pi_{c_1, \dots, c_n} (\sigma_{\text{predicate}} (S)) \equiv \sigma_{\text{predicate}} (\pi_{c_1, \dots, c_n} (S))$ iff predicate only uses c_1, \dots, c_n
- Commutability between selection & join (predicate pushdown)
 - $\sigma_{\text{predicate}} (R \text{ join } S) \equiv (\sigma_{\text{predicate}} (R)) \text{ join } S$ iff predicate only uses attributes from R
- Commutability between projection & join
 - $\pi_{c_1, \dots, c_n} (R \text{ join}_{cr=cs} S) \equiv (\pi_{c_1, \dots, c_n, cr} (R)) \text{ join}_{cr=cs} S$

Example: Using Equivalences



Cost Estimation

- Obvious inefficient plans are pruned during enumeration. Eg. Predicate pushdown etc.
- For each plan considered,
 - Must *estimate cost* of each operation in plan tree.
 - Depends on input cardinalities.
 - We've already discussed how to estimate the cost of operations (sequential scan, index scan, joins, etc.)
 - Must also *estimate size of result* for each operation in tree!
 - Use information about the input relations.
 - For selections and joins, assume independence of predicates.

Example: Predicate Pushdown

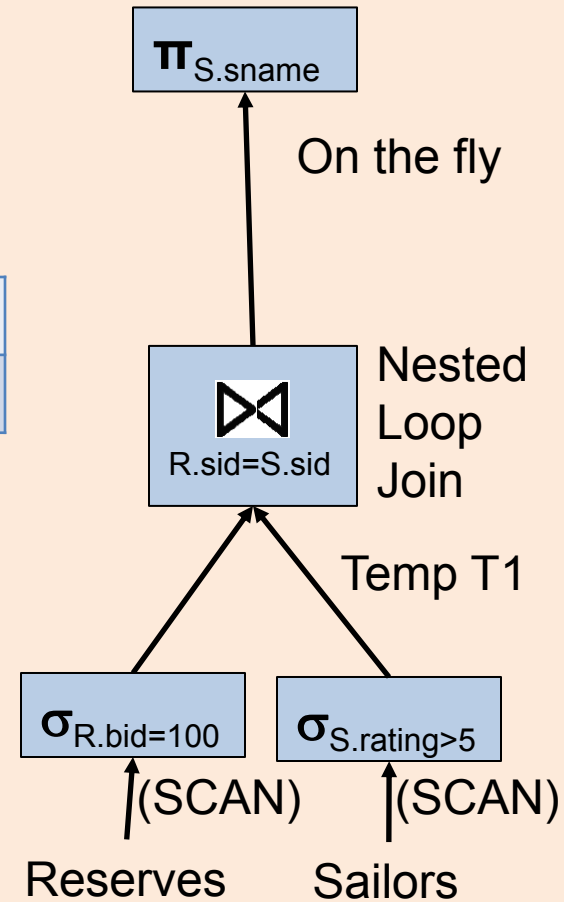
SELECT S.sname
FROM Reserves R, Sailors S
WHERE R.sid=S.sid **AND** R.bid=100 **AND** S.rating>5

10%

50%

Reserves	40 bytes/tuple	100 tuples/page	1000 pages
Sailors	50 bytes/tuple	80 tuples/page	500 pages

- Nested Loop Join requires materializing the inner table as T1.
- With 50% selectivity, T1 has 250 pages
- With 10% selectivity, outer “table” in join has 10K tuples
- Disk accesses for scans = 1000 + 500
- Writing T1 = 250
- NLJoin = 10K * 250
- Total disk access = 2500.175 K (worst case)



What happens if we make the left leg the inner table of the join ?

Example: Sort Merge Join

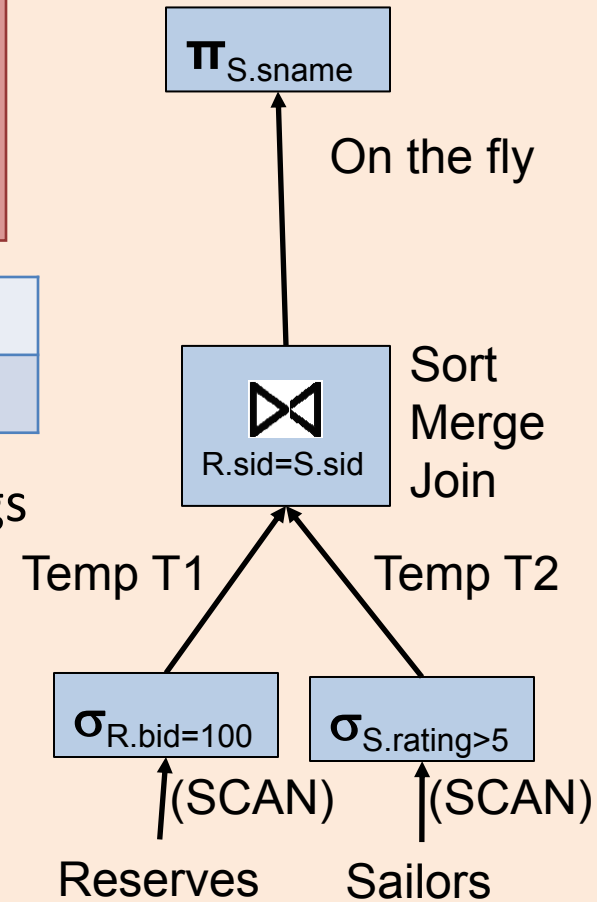
SELECT S.sname
FROM Reserves R, Sailors S
WHERE R.sid=S.sid **AND** R.bid=100 **AND** S.rating>5

10%

50%

Reserves	40 bytes/tuple	100 tuples/page	1000 pages
Sailors	50 bytes/tuple	80 tuples/page	500 pages

- Sort Merge Join requires materializing both legs for sorting.
- With 10% selectivity, T1 has 100 pages
- With 50% selectivity, T2 has 250 pages
- Disk accesses for scans = 1000 + 500
- Writing T1 & T2 = 100 + 250
- Sort Merge Join = $100 \log 100 + 250 \log 250 + 100 + 250$ (assume 10 way merge sort)
- Total disk access = 52.8 K



What happens if we make the left leg the inner table of the join ?

Example: Index Nested Loop Join

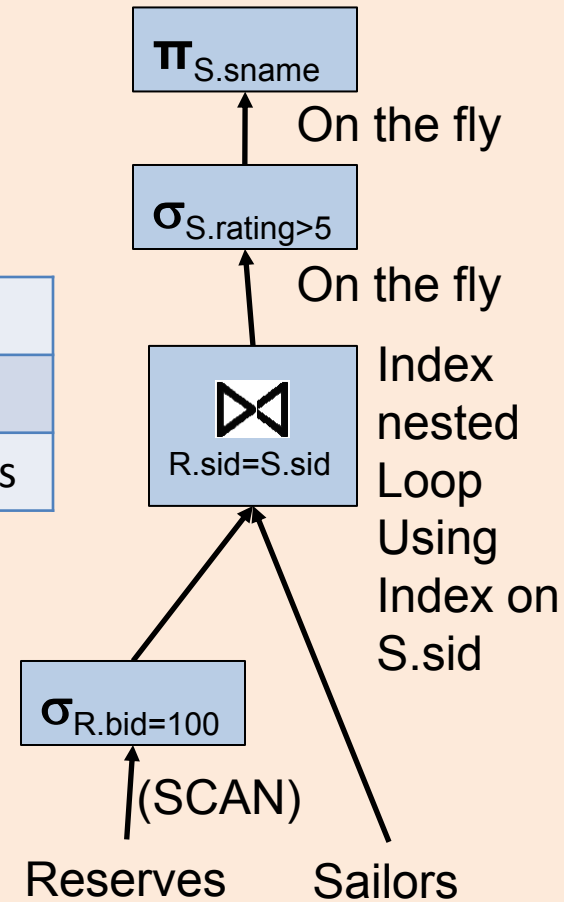
SELECT S.sname
FROM Reserves R, Sailors S
WHERE R.sid=S.sid **AND** R.bid=100 **AND** S.rating>5

10%

50%

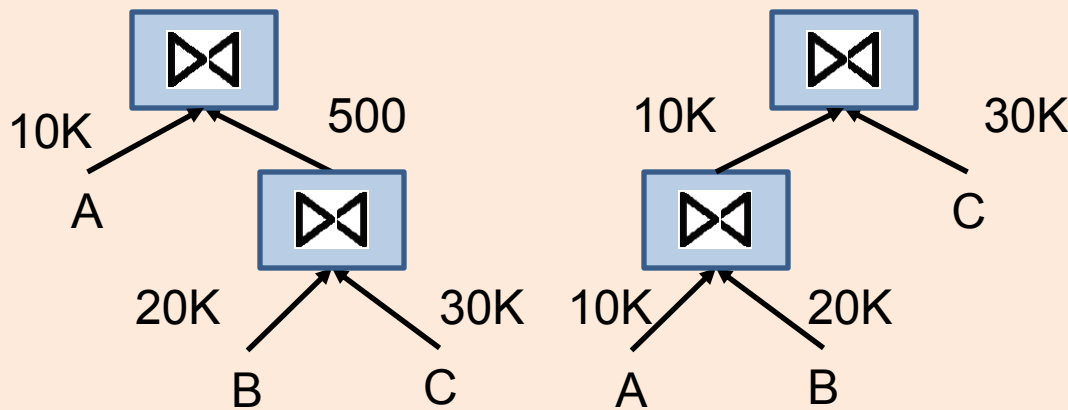
Reserves	40 bytes/tuple	100 tuples/page	1000 pages
Sailors	50 bytes/tuple	80 tuples/page	500 pages
Index(S.sid)			200 leaf pages

- With 10% selectivity, selection on R has 10K tuples
- Disk accesses for scan = 1000
- Index Nested Loop Join = $10K * (1 + \log_{10} 200) = 33K$
- Total disk access = 34 K



What happens if we make the left leg the inner table of the join ?

Join Ordering



Relations	Tuples	Pages
A	10K	1000
B	20K	2000
C	30K	3000
A join B	10K	1000
B join C	1K	100

- Independent of what join algorithm is chosen, the order in which joins are performed affects the performance.
- Rule of thumb: do the most “selective” join first
- In practice, left deep trees (eg. the right one above) are preferred --- why ?

How to estimate the selectivity & cardinality ?

$\sigma_{col=value}$

- Arbitrary constant 10%
- $1 / \text{Number of distinct values in the column}$
- $1 / \text{Number of keys in Index(col)}$

$\sigma_{col>value}$

- Arbitrary constant of 50% if non numeric
- $(\text{High Key} - \text{value}) / (\text{High Key} - \text{Low Key})$

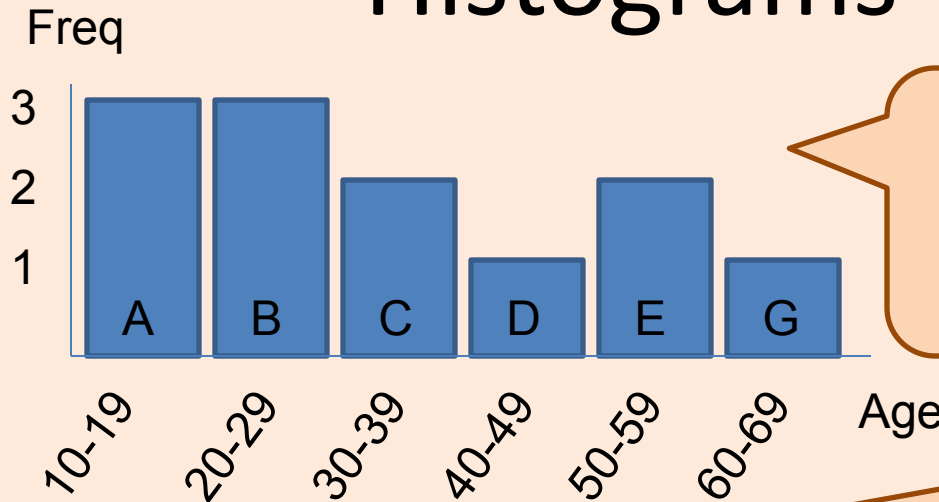
$\sigma_{R.col=S.col}$

- Join result size
- Arbitrary constant 10%
- $1 / \text{MAX}(\text{Nkeys}(\text{Index}(\text{R.col})), \text{Nkeys}(\text{Index}(\text{S.col})))$

Can we do better ?

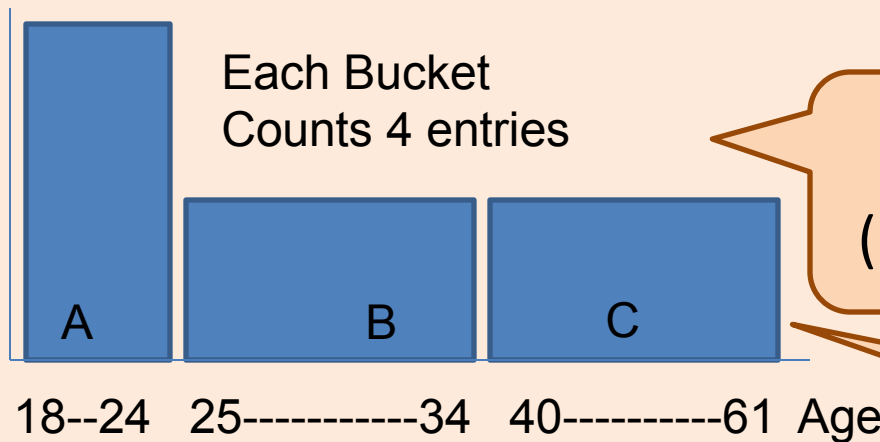
Histograms

Age
18
18
19
24
25
29
30
34
40
50
58
61



$$\sigma_{\text{age} > 45} : 0.5 * f(D) + f(E) + f(G)$$

Equi-width buckets



$$\sigma_{\text{age} > 30} : (34-30)/(34-25) * B + C$$

Equi-depth buckets

Statistics Collection in DBMS

- Page size
- Data Statistics:
 - Record size -> number of records per data page
 - Cardinality of relations (including temporary tables)
 - Selectivity of selection operator on different columns of a relation
- (Tree) Index Statistics
 - number of leaf pages, index entries
 - Height
- Statistics collection is user triggered
 - DB2: RUNSTATS ON TABLE mytable AND INDEXES ALL

What about the parallel/distributed case?

- QEP enumeration/rewrite
 - Main “trick” is expressing a horizontally fragmented table as a union of fragments in RA
 - Push the union up. Conversely push the σ, π, \times down.
 - Eliminate sub-trees that return empty results.
- Cost estimation takes into account communication costs.

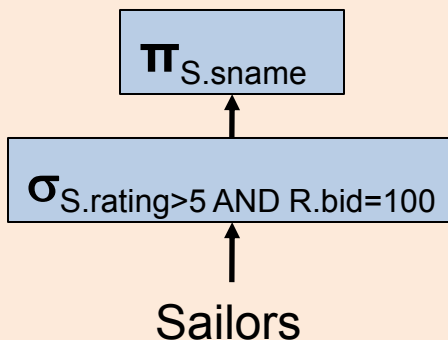
```

SELECT S.sname
FROM Sailors S
WHERE S.rating>5
    
```

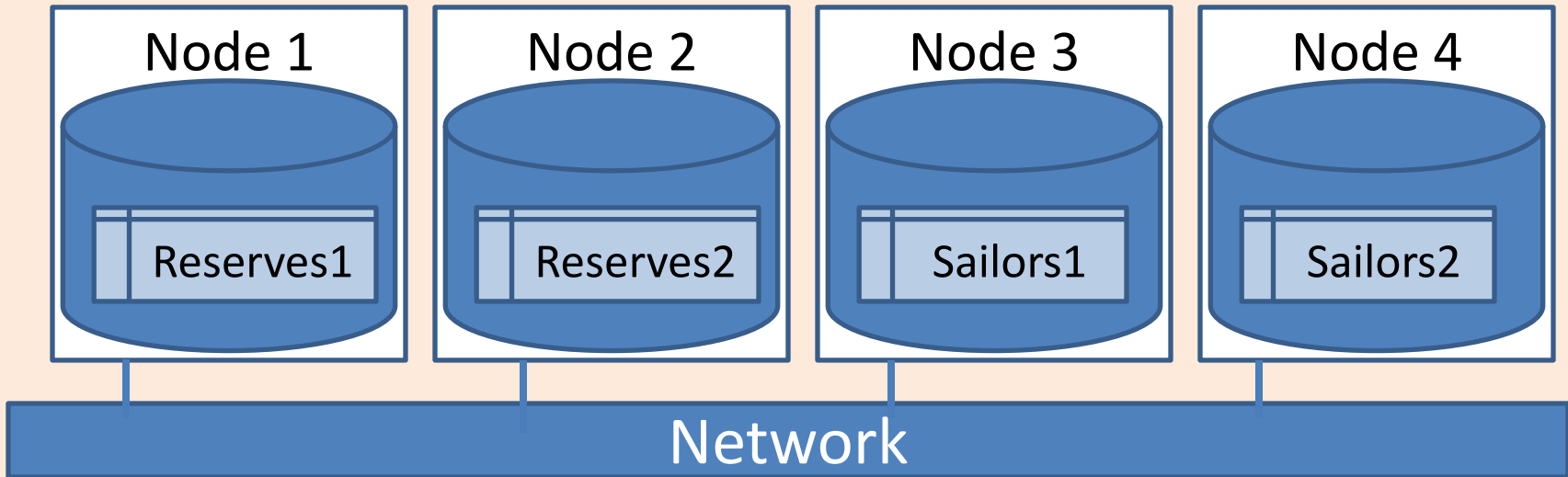
Sailors partitioned by ranges [0,5),[5,10] on rating

$$\begin{aligned}
 & \pi_{S.sname} (\sigma_{S.rating>5} S) \\
 &= \pi_{S.sname} (\sigma_{S.rating>5} (S1 \cup S2)) \\
 &= \pi_{S.sname} ((\sigma_{S.rating>5} S1) \cup (\sigma_{S.rating>5} S2)) \\
 &= (\pi_{S.sname} \sigma_{S.rating>5} S1) \cup (\pi_{S.sname} \sigma_{S.rating>5} S2)
 \end{aligned}$$

S1's range is [0,5), so selection is empty!



Distributed Multi-table Query



$$\begin{aligned} R \text{ join } S &= \sigma_{R.sid=S.sid} (R \times S) \\ &= \sigma_{R.sid=S.sid} ((R1 \cup R2) \times (S1 \cup S2)) \\ &= \sigma_{R.sid=S.sid} ((R1 \times S1) \cup (R1 \times S2) \cup (R2 \times S1) \cup (R2 \times S2)) \\ &= \sigma_{R.sid=S.sid} (R1 \times S1) \cup \sigma_{R.sid=S.sid} (R1 \times S2) \\ &\quad \cup \sigma_{R.sid=S.sid} (R2 \times S1) \cup \sigma_{R.sid=S.sid} (R2 \times S2) \\ &= (R1 \text{ join } S1) \cup (R1 \text{ join } S2) \cup (R2 \text{ join } S1) \cup (R2 \text{ join } S2) \end{aligned}$$

Equivalent to a union of joins over each pair of fragments