

### Costs for Hash Join

- Assuming all tuples of both *R* and *S* can be place on main memory
  - Consider disk base hash join later
- We have to consider the following costs
  - Applying Hash Function:  $\{R\} + \{S\}$
  - − Building the Hash Table: {R}
  - Comparisons: {S}

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### Hash Join with Disk Accesses

- Three Algorithms
  - Simple Hash Join
  - GRACE Hash Join
  - Hybrid Hash Join
- Assumptions for cost estimation
  - Memory size for hash table: |M|
  - The distribution of attribute value is flat
  - -|R|>|S|

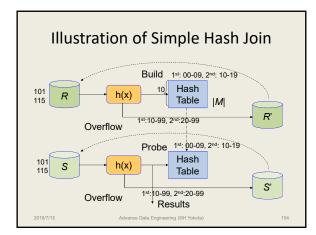
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### Simple Hash Join

- Restore tuples that cannot be placed in main memory hash table
  - Repeat until no overflow
  - Both R and S (because of the identical hash table)
- The number of repeat (Expectation):
  - $L = |R|/|M| (\ge 1)$

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## **Example of Overflow Control**

- Suppose an attribute storing integer value
- Cardinality of R is 100
- The hash table  $|\,M\,|$  can keep 10 tuples of R
- Apply hash function h(X)=mod(X,100)
  - It generates two-digit numbers
  - Lower digit is used to choose an entry of the hash table
  - Higher digit is used to decide overflow tuples
- The first iteration: tuples of R, having higher digit of hash value is 0, are used to build the hash table, and others become overflow
- The second iteration: tuples having higher digit  $\neq 1$  becomes overflow
- And so on
- Iteration count becomes 10 or more

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# Cost for Simple Hash Join (1)

- In the first loop
  - Applying Hash Function: {R} + {S}
  - Comparison: {S} / L
  - Disk I/O for read: |R| + |S|
  - Disk I/O for write: |R| |M| + |S| |S| / L
  - or (|R| + |S|) (|R| + |S|) / L
  - from |M| = |R| / L

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## Cost for Simple Hash Join (2)

- In the *i*-th loop (i = 2, ..., L)
  - Applying Hash Function:
    - $\{R\}$  (i-1) ×  $\{R\}$  / L +  $\{S\}$  (i-1) ×  $\{S\}$  / L
  - Comparison: {S} / L
  - Disk I/O for read:
  - $(|R| + |S|) (i 1) \times (|R| + |S|) / L$
  - Disk I/O for write:

 $(|R| + |S|) - i \times (|R| + |S|) / L$ 

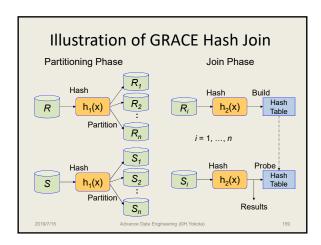
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### Cost for Simple Hash Join (3)

- Summation from the first to L-th (last) loop
  - Applying Hash Function:
    - $L \times (\{R\} + \{S\}) (L \times (L-1)/2 \times (\{R\}/L + \{S\}/L))$
  - $= (L + 1) / 2 \times (\{R\} + \{S\})$
  - $= (|R| + |M|) \times (\{R\} + \{S\}) / 2|M|$
  - Comparison:  $(\{S\} / L) \times L = \{S\}$
  - Disk I/O :
  - $2L \times (|R|+|S|) (L \times (L+1)) / 2$
  - + L x (L-1) / 2) x (|R|+|S|) / L
  - $= L \times (|R| + |S|)$
  - $= (|R| \times (|R| + |S|)) / |M|$

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## Cost for GRACE Hash Join

- Disk I/O: 3 (|R| + |S|)
  - Read whole |R| and |S| and write them in the partition phase, and read them again in the join phase
- Comparison :  $\Sigma S_i = (\{S\} / L) \times L = \{S\}$ 
  - Assume Hash Join in Join Phase
- Applying Hash Function:
  - Partitioning Phase:  $\{R\} + \{S\}$ - Join Phase:  $\Sigma(\{R_i\} + \{S_i\}) = \{R\} + \{S\}$
  - Whole: 2 ( $\{R\}$  +  $\{S\}$ )

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### Illustration of Hybrid Hash Join Partitioning Phase Join Phase Build Hash Hash Build $h_1(x)$ $R_i$ $h_2(x)$ Partition $R_n$ $i=2,\,\ldots,\,n$ Results Hash Probe $h_1(x)$ Partition Results

# Cost for Hybrid Hash Join

- Disk I/O :
  - $\begin{array}{l} -3 \left( |R| + |S| \right) 2 \left( |R| + |S| \right) / L \\ = \left( 3 2 |M| / |R| \right) \times \left( |R| + |S| \right) \end{array}$
- Comparison and Applying Hash Function
  - The same as GRACE Hash Join

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## Comparison of Hash Join Algorithm

|        | Disk I/O                        | Applying<br>Hash                    | Comparison |
|--------|---------------------------------|-------------------------------------|------------|
| Simple | $\frac{ R }{ M }( R + S )$      | $\frac{ R + M }{2 M }(\{R\}+\{S\})$ | {S}        |
| GRACE  | 3 ( R + S )                     | 2 ({R}+{S})                         | {S}        |
| Hybrid | $(3-2\frac{ M }{ R })( R + S )$ | 2 ({R}+{S})                         | {S}        |

# Optimization of Join (1)

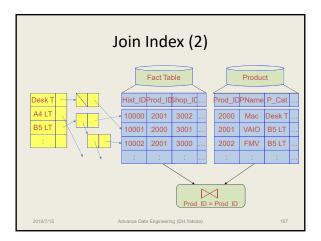
- Star Query (Oracle 7)
  - Derive Cartesian Products among Dimension Tables
    - to avoid handling the huge Fact Table
  - Tradeoff between the cost for Cartesian Product and that for handling the Fact Table
  - The dimension tables can be filtered by conditions in advance
    - e.g. A4 Laptop, Aug., etc

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### Optimization of Join (2) · Bit Vector Filtering Bloom Filtering Merge Join - Make each entry of R" the Hash Table 1 bit Probe Vector S' Only Existence (Allow R' Filtering S' - Bit Vector can be place Sort Bit Map Build Vector S' in main memory R 1 S' - This filtering can apply Build Vector $R \longrightarrow Probe Vector R$ both sort merge and R | Hash Bit Map S | Filtering hash join Scan R Scan S

# Join Index (1)

- Assuming join attributes for relations R and S are R.A and S.B, respectively, an index from the other attribute of R.C of the relation R is a Join Index.
  - Structure: Inverted File or B+-tree
  - Collisions for one entry: Inverted List or in Inverted File / B-tree



# 

### Bitmap Index

- Toggle 0 or 1 by existence of a value of the attribute
- When the variety of values is small, it provides good space efficiency
  - Bitmap can be place in main memory
  - Bitmap can be calculated as an array
    - AND/OR operation can be used for filtering for (i = 0; i < len(B1); i++) B3[i] = B1[i] & B2[i];
- On the other hand, large variety of attribute values make the space efficiency worse.
  - combination with value index by segmentation is proposed

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### An Example of Bitmap Index Fact Table Prod ID Shop\_ID Bitmap Index Results Bitmap Index Hist\_IDProd\_IDShop\_ID 2001 10001 2000 3001 10002 2001 3000 10003 2020 3002 3010 10004 2003 10002 2001 10003 2002 3011

### Semi-Join Operation • A kind of eq-join, but attributes in the result relation belong to only one relation Product ID Product Type Product ID Stocks P001 Laptop PC 1,500 P001 10 P002 Laptop PC 2,000 P003 5 Desktop PC P003 2,000 Product ID Product Type Price P001 1,500 Laptop PC 2,000 P003 Desktop PC

| • Similar t | nti-Semi-Join,<br>o semi-join,<br>oes not muc | but ge    | ener         | ati          | ng tuple  |        |  |  |
|-------------|---|-----------|--------------|--------------|-----------|--------|--|--|
| Product ID  | Product Type                                  | Price     |              |              | roduct ID | Stocks |  |  |
| P001        | Laptop PC                                     | 1,500     |              | P001<br>P003 |           | 10     |  |  |
| P002        | Laptop PC                                     | 2,000     | <del> </del> |              |           | 5      |  |  |
| P003        | Desktop PC                                    | 2,000     | Į l          |              |           | 3      |  |  |
|             |   |           |              |              |           |        |  |  |
|             | Product ID Product                            |           | uct Ty       | ре           | Price     |        |  |  |
|             | P002  | Laptop PC |              | PC 2,000     |           |        |  |  |
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### Costs of Semi-Join & Anti-Semi-Join

- Cost of Semi-Join are basically same as ordinary join operations
  - For the case of hash join, the size of hash table can be reduced
- Implementation of anti-semi-join is also similar to semijoin operation except outputting unmuched tuples instead of muched tuples
  - The cost of anti-semi-join is equal to semi join
- These operations are related to set operations
- Semi-Join is also related to distributed database

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# Implementation of Set Operations

- Intersection  $(R \cap S)$ 
  - Apply the semi-join targeting all attributes as conditions
  - Costs is equal to one for a join operation
- Difference (R S)
  - Apply the anti-semi-join targeting all attributes
  - Costs is also equal to one for a join operation
- Union  $(R \cup S)$ 
  - Apply difference operation to derive (R-S) and concatenate (R-S) with S
  - Or concatenate R with S and eliminate duplications

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### Division ÷ Division produce a relation Product ID Salesman that consists of the set of P001 John S tuples from R defined over P001 Alice Product ID the attribute C that match P001 Bill P001 the combination of every P002 John P002 tuple in S, where C is the P002 Bill P003 set of attributes that are in P003 John R but not in S P003 Bill P004 Alice P004 $R[ProductID] \div S[ProductID]$ Salesman Salesman who sold all products Bill listed in relation S John

### Implementation of $R[X] \div S[X]$ (1) Sort Base Direct Method Sort R by X as minor and C as major Check X of R with S from top of the same value C (X) R (C)minor maior S Product ID Salesman Product ID Salesman Product ID P001 Alice P001 P004 Alice P002 P001 Sort P003 John P002 P002 Bill P003 results P003 John P004 Bill P001 P003 Bill John P002 Bill P004 Alice John John P004 Bill John

### Implementation of $R[X] \div S[X]$ (2) Hash Base Direct Method Prepare two hash functions h1(X) and h2(C) h1(X) is used to derive entry # in the indicated h2(C) and set 1 P002 Product ID Salesman P001 John 2 P001 P001 Alice 3 P001 Bill P002 John P002 P001 P003 P002 Bill 1 1 John P003 John Alice 1 P003 Bill P004 Alice Bill 1 1 1 P004

| Implementation of $R[X] \div S[X]$ (3)  |          |                          |            |   |            |          |       |   |  |
|---|----------|--------------------------|------------|---|------------|----------|-------|---|--|
| Aggregate Function based Indirect Method     Apply semi-join in advance, then count tuples for each group of C     If any group have count as same as the cardinality of S, then output |          |                          |            |   |            |          |       |   |  |
| Product ID Salesman   |          | 1                        | Product ID |   | Product ID |          |       |   |  |
| Productib   | Salesman |                          | Product ID |   | Productib  | Salesman |       |   |  |
| P001  | John     |                          | P001       |   | P001       | John     | Labor | 3 |  |
| P001  | Alice    | $  \times  $             | P002       |   | P001       | Alice    | John  | _ |  |
| P001  | Bill     |                          | P003       | 7 | P001       | Bill     | Alice | 1 |  |
| P002  | John     |                          |            |   | P002       | John     | Bill  | 3 |  |
| P002  | Bill     |                          |            |   | P002       | Bill     | /     |   |  |
| P003  | John     |                          |            |   | P003       | John     | /     |   |  |
| P003  | Bill     |                          |            |   | P003       | Bill     | /     |   |  |
| P004  | Alice    | Quiz:                    |            |   |            |          |       |   |  |
| P004  | Bill     | Consider an SQL sentence |            |   |            |          |       |   |  |
| 20197/15 to derive the results  |          |                          |            |   |            |          |       |   |  |

# Cost for Group-by and Aggregation Operations (1) • Nested Loop base - Scan all tuples to search tuples in the same group, and then calculate COUNT, SUM, AVG, MAX, MIN • $(\{R\}-1\}+(\{R\}-2\}+...+1$ • Comparison : $\{R\}\{\{R\}-1\}/2$ - Write the result into disk at the last • Disk I/O: |R|(|R|+1)/2

### Cost for Group-by and Aggregation Operations (2)

- · Sort base
  - At first sort all tuples and scan from the top
    - · Assume Merge Sort
  - Cost for sort and one scan at the last
    - Comparison : {R}(log{R}+1)
  - Disk I/O:  $|R|(2 \log |R|+1)$
- Hash base
  - Divide tuples by a hash function
    - · Apply the aggregate function for each hash bucket
  - If there is no hash collision
    - Applying Hash Function, Comparison: {R}
       Disk I/O: |R|

# **Indexing for Aggregation Functions**

- Bit-Sliced Index
  - Divide each bit of binary expression of an integer
    - A list of each bit in tuples is treated as a bitmap index
    - Each bit-slice can be placed on main memory
    - Calculate SUM or AVG for each bit-slice and summarize
  - Adopted by Sybase IQ, CCA Model 204

## **Indexing for Aggregation Functions**

- Projection Index
  - Derive an attribute (Projection), and access by its location.
    - Reduce Disk I/O by reducing amount of data
  - $-\,\mbox{Suited}$  for complex aggregate functions on the
  - Derive SUM for the results of some calculation
  - Adopted by Sybase IQ

### Comparison on the AF Indices Aggregate Value-list Bit-Sliced Projection Slow Max, Min Best Slow SUM, AVG Not Bad Best Good SUM(A1\*(1-A2)) Very Slow Very Slow Best Narrow Best Good Good Range Wide Not Bad Good Best Range

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