

# Lecture 10: Decision support, OLAP

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Reading: RG25, [WOS04, sec. 1+2],  
[DSTW03, sec. 1+2+3.0]

# Today's lecture

- Multi-dimensional data and OLAP.
- Bitmap indexing.
- Materialized views: Use and maintenance.
  
- Exercises.
- "On producing join results early".
- Some other uses of sampling.

# Background

- Data of an organization is often a gold mine!
  - Data can identify useful patterns that can be used to form a proper business strategy.
- Two main approaches:
  - Data mining: Automated “mining” of patterns.
  - OLAP: Interactive investigation of data. (This lecture.)
- OLAP = On-Line Analytic Processing  
OLTP = On-Line Transaction Processing

# Multi-dimensional data model

- Tuples with  $d$  attributes can be viewed as points in a  $d$ -dimensional "cube".
- **Example:** (DBT, 'John Doe', 13) can be viewed as the point with coordinates DBT, 'John Doe', and 13 along the dimensions Course, Name, and Grade.
- Natural view of data, specifically for *aggregation* queries. Dimensions often have several *granularities*.
- **Example:** We may want to compute the average grade of all SDT courses.

# Sample OLAP query (in SQL)

```
SELECT SUM(S.sales)
FROM Sales S, Times T, Locations L
WHERE S.timeid=T.timeid AND S.locid=L.locid
GROUP BY T.year, L.state
```

```
SELECT SUM(S.sales)
FROM Sales S, Times T
WHERE S.timeid=T.timeid
GROUP BY T.year
```

```
SELECT SUM(S.sales)
FROM Sales S, Location L
WHERE S.timeid=L.timeid
GROUP BY L.state
```

Get fine-grained  
aggregate data

Get dimension  
"coordinates"  
+ aggregates



# SQL:1999 CUBE operator

- Similar to GROUP BY, but includes groupings according to *all subsets* of the given attributes.

- **Example:**

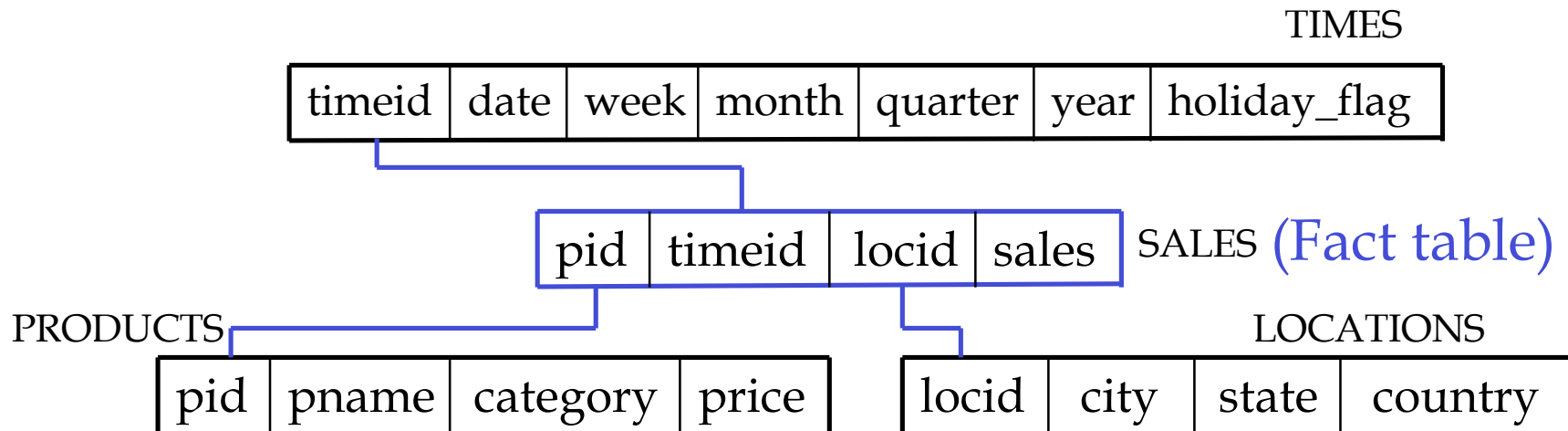
```
SELECT year, state, SUM(sales)
FROM Salesview
GROUP BY CUBE (year, state)
```

returns tuples of the form:

- (1999, Iowa, 29438)
- (1999, NULL, 314974)
- (NULL, Iowa, 213891)
- (NULL, NULL, 3217919)

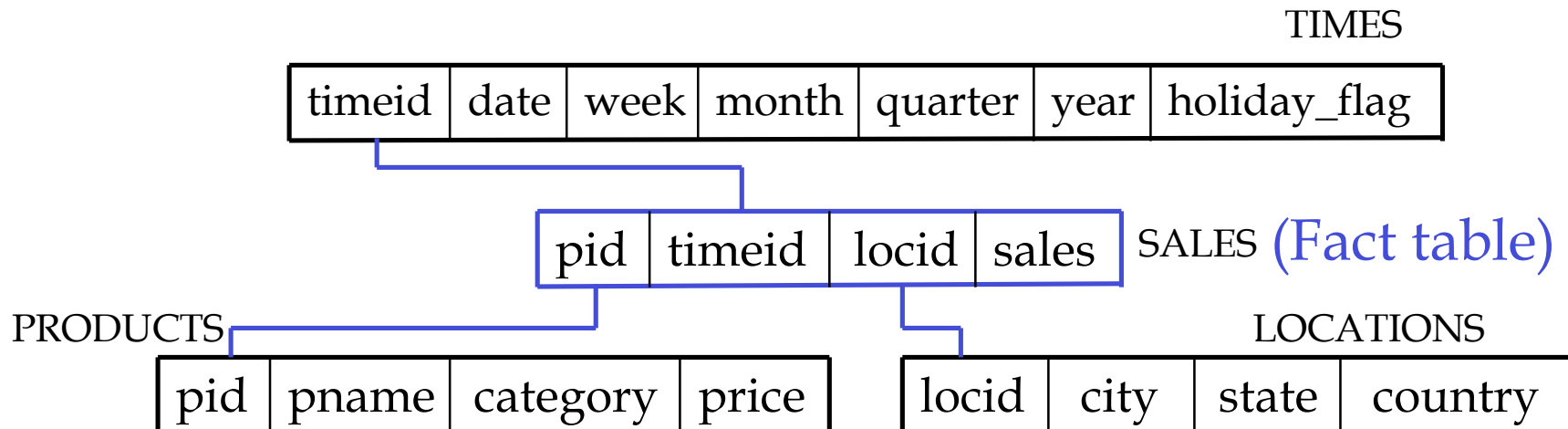
# Relational set-up of OLAP

- **Star schema:** A "fact table", plus a number of "dimension tables" whose keys are foreign keys of the fact table.
- Example from RG's slides:



# Problem session

- How would you efficiently answer the query: *“Find the total sales of raincoats for each location last year.”*
- You may assume suitable indexes.





# Queries on star schemas

## Special considerations:

- Number of relations can be large - join size estimation is difficult.
- Dimension tables are often small, especially when considering only the tuples matching a select operation.
- Often data can be assumed to be *static* (a snapshot of the operational data). This means that we have the option to *precompute*.



# Indexing low cardinality attributes

- Suppose there are only 4 different locations in our previous example.
- Then we may represent the locations of  $N$  fact table tuples using only  $2N$  bits.
- However, a (clustered) index on location seems to require at least  $N \log N$  bits.
- Or??

# Basic bitmap index

- For each possible value (of "location") and each tuple, store a bit that is 1 iff the tuple contains the given value.
- Store these bits **ordered by column**.
- In the context of this example, the bitmap index is a **join index** that can be used when computing joins.
- How?

L1?	L2?	L3?	L4?	RID
0	0	1	0	1
1	0	0	0	2
1	0	0	0	3
0	1	0	0	4
0	0	0	1	5

# Problem session

- How much can **at most** be gained by using bitmap indexes to do a star join (with a selection on each dimension table), compared to using B-tree indexes?
- Are there cases where there is no gain?
  - For some data?
  - Independent of data?

# Compressed bitmap indexes

- If there are many possible values for an attribute (it has "high cardinality"), basic bitmap indexing is not space efficient (nor time efficient).
- **Observation:** A column will have few 1s, on average. It should be possible to "compress" long sequences of 0s.
- **How to compress?** Usual compression algorithms consume too much computation time. Need simpler approach.



# Word-aligned hybrid (WAH) coding

[WOS04]

- In a nutshell:
  - Split the bitmap B into pieces of 31 bits.
  - A 32-bit word in the encoding contains one of the following, depending on the value of its first bit:
    - A number specifying the length of an interval of bits where all bits of B are zeros.
    - A piece of B (31 bits).
  - The conjunction ("AND") or disjunction ("OR") of two compressed bitmaps can be computed in linear time ( $O(1)$  ops/word).



# WAH analysis

- Let  $N$  be the number of rows of the indexed relation, and  $c$  the cardinality of the indexed attribute.
- At most  $N$  WAH words will encode a piece of the bitmap.
- Reasonable assumption:
  - All (or most) gaps between consecutive 1s can be encoded using 31 bits.
  - Thus, at most  $N+c$  gaps.
- Total space usage:  $2N+c$  words.
- Compares favorably to B-trees.

# WAH experiments

- Implemented in the "FastBit" system.
- Compared against compressed bitmap index of Oracle: Considerable speedup due to simpler program.
- Compared against not using a bitmap index (MySQL): Speedup several orders of magnitude.



# De-confuser

- The book talks about join indexes in two distinct senses:
  - Precomputing the join result (“basic join index”).
  - Precomputing *projections* of the join result onto tuples of pairs of relations.
- In both cases, compact row IDs (rids) are used to represent the tuples forming each result tuple.
- It is for the latter that we can use bitmap indexing with advantage.

## Next: Materialized views

- An SQL *view* is similar to a macro.
- Example:

```
CREATE VIEW MyView AS
SELECT *
FROM     Sales S, Times T, Locations L
WHERE    S.timeid=T.timeid AND
         S.locid=L.locid
```
- A query on MyView is transformed into a query that performs the join of Sales, Times, and Locations.
- In contrast, a **materialized view** *physically* stores the query result. Additionally: can be indexed!

# Using a materialized view

1. DBA grants permission:

```
GRANT CREATE MATERIALIZED VIEW TO hr
```

2. Materialized view is created:

```
CREATE MATERIALIZED VIEW SalaryByLocation AS  
SELECT location_id, country_id, SUM(salary) AS s  
FROM Employees NATURAL JOIN Departments  
      NATURAL JOIN Locations  
GROUP BY location_id, country_id
```

3. Materialized view is used:

```
SELECT country_id, SUM(salary) AS salary  
FROM SalaryByLocation  
GROUP BY country_id
```

Factor 10 faster than direct query on  
Oracle XEs example DB.



## Automatically using mat. views

- Suppose a user does not know about the materialized view and writes directly

```
SELECT location_id, country_id, SUM(salary) AS s
FROM Employees NATURAL JOIN Departments
      NATURAL JOIN Locations
GROUP BY country_id
```

- A smart DBMS will realize that this can be rewritten to a query on the materialized view. (Disabled in Oracle XE...)
- Rewrite capability is a key technique in relational OLAP systems.

# "Refreshing" a materialized view

- Any change to the underlying tables may give rise to a change in the materialized view. There are at least three options:
  - Update for every change ("ON COMMIT")
  - Update only on request ("ON DEMAND")
  - Update when the view is accessed ("lazy")
- RG describes a way of refreshing where recomputing the defining query is often not necessary ("FAST"). ( (board) )

# Exercises

We look at two exercises from RG:

- 25.8
- 25.10

## Schema:

Locations (`locid, city, state, country`)

Products (`pid, pname, category, price`)

Sales (`pid, timeid, locid, sales`)

## Next: Early join results

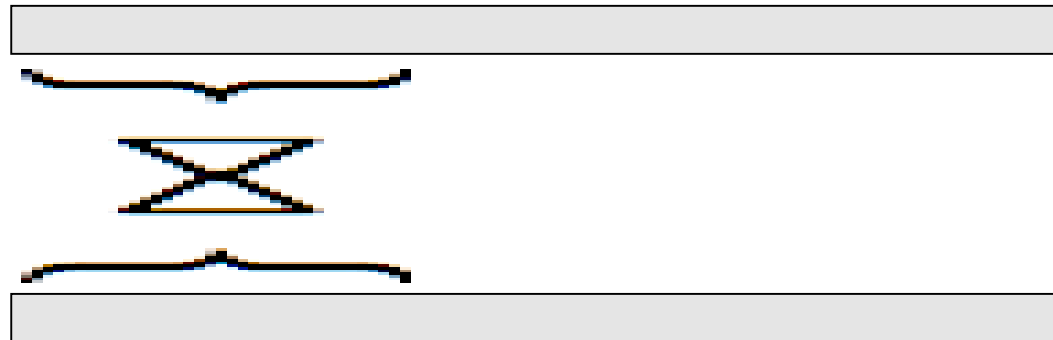
[DSTW03]

- Even with suitable materialized views in place, we may need to do some joins.
- **OLAP goal:** Get at least *part* of the result early.
- We consider the simplest case: Joining two relations R and S
- Basic idea: First report  $R' \bowtie S'$ , where  $R'$  and  $S'$  are samples (subsets) of R and S.
- Assumption: We get a "random-like" sample by picking the first k tuples.



# First approach

- Join samples of size  $1, 2, 4, 8, \dots, N$ . (TPMMS)
- Report join tuples as soon as they are seen - remember to filter away tuples that were previously reported.



- **Problem session:**
  - How many tuples can we expect after joining the samples of size  $k$ ?
  - What is the total I/O cost of this?



# Progressive mergesort join

- Same idea (increasing sample sizes).
- New: [DSTW03]
  - Re-use as much work as possible
  - Avoid explicit filtering

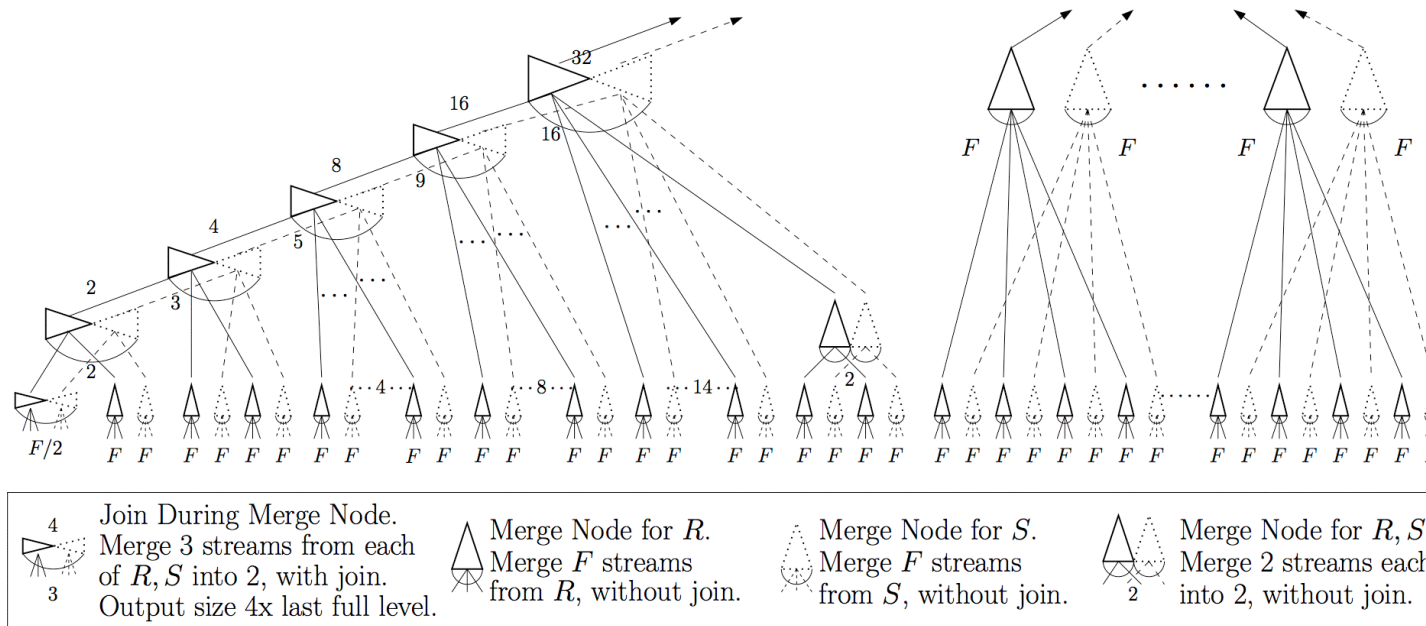


Figure 1: One full level of Progressive Mergesort Join to the next ( $F = 64$ ).

# On-line aggregation

- For aggregates like sums and averages, the result on a sample can be used to estimate the result on all data.
- Same principle as used in opinion polls!
- Can give statistical guarantees on an answer, e.g. "Answer is  $3200 \pm 180$  with 95% probability".
- The longer the query runs, the smaller the uncertainty gets.
- Possibly ok to terminate before precise answer is known.

# "Top K" queries

- Suppose, in a given query result, we are only interested in the K tuples with the largest values of attribute A.
- In Oracle:

```
SELECT *  
FROM (SELECT ... ORDER BY A DESC)  
WHERE rownum<=K.
```
- If the DBMS knew the "cutoff" value for A, we could add this as a condition, possibly reducing dramatically the amount of data to be considered.
- Sampling approach: Estimate (conservatively) the right cutoff based on the sample.

