# Lecture 13 Optimizations

## **Summary – last week**



- R+Tree
  - Removes the overlaps in R Tree
- Bitmap indexes
  - Great for indexing tables with set-like attributes e.g.,
     Gender: Male/Female
  - Operations are efficient and easy to implement (directly supported by hardware)



## This week

### **Optimization**

I. Partitioning



## Why Partitioning?

## **Partitioning**

 Breaking the data into several physical units that can be handled separately

- Granularity and partitioning are key to efficient implementation of a warehouse
- The question is not whether to use partitioning but how to do it

- Why partitioning?
  - Flexibility in managing data
  - Smaller physical units allow
    - Easy restructuring e.g. adding weekly classification level
    - Cheap indexing
    - Sequential scans, if needed
    - Easy reorganization
    - Easy recovery
    - Easy monitoring



- In DWs, partitioning is done to improve:
  - Business query performance, i.e., minimize the amount of data to scan
  - Data availability, e.g., back-up/restores can run at the partition level
  - Database administration, e.g., adding new columns to a table, archiving data, recreating indexes, loading tables

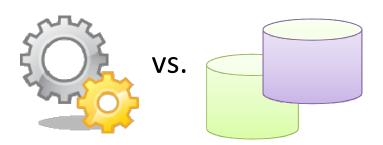
- Possible approaches:
  - Data partitioning
     where data is usually partitioned by
    - Date
    - Line of business
    - Geography
    - Organizational unit
    - Combinations of these factors

#### - Hardware partitioning

 Hardware partitioning makes data available to different processing nodes by ensuring that sub-processes are capable of running on the different nodes



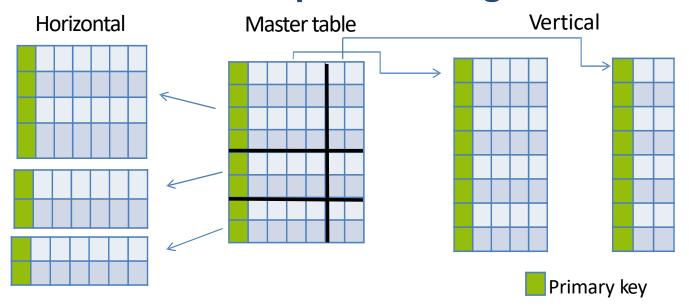
- At which level should data partitioning occur?
  - Possibilities are
    - on application
    - or DBMS level



- Partitioning on DBMS level is clear, but it also makes sense to partition at application level
  - E.g., allows different definitions for each year
    - Important since DWs spans many years and as business evolves DWs change, too
    - Think for instance about changing tax laws

## **Data Partitioning**

- Data partitioning, involves:
  - Splitting out the rows of a table into multiple tables
     i.e., horizontal partitioning
  - Splitting out the columns of a table into multiple tables i.e., vertical partitioning



Which partitioning type is good tax laws changing every year?

## Data Partitioning (cont'd.)

- Horizontal partitioning
  - The set of tuples of a table is split among disjoint table parts
  - Definition: A set of Relations  $\{R_1,...,R_n\}$  represent a **horizontal partitioning** of Master-Relation R if and only if  $R_i \subseteq R$ ,  $R_i \cap R_j = \emptyset$  and  $R = U_i R_i$ , for  $1 \le i, j \le n$
  - According to the partitioning procedure we have different horizontal partitioning solutions
    - Range partitioning, List partitioning and Hash partitioning

## **Horizontal Partitioning**

- Range Partitioning
  - Selects a partition by determining if the partitioning key is inside a certain range
  - A partition can be represented as a restriction on the master-relation
    - $R_i = \sigma_{Pi}(R)$ , where  $P_i$  is the partitioning predicate. The partitioning predicate can involve more attributes
      - $-P_1$ :Country = 'Germany' and Year = 2009
      - − P₂:Country = 'Germany' and Year < 2009</p>
      - P<sub>3</sub>:Country ≠ 'Germany'

## Horizontal Partitioning (cont'd.)

#### List Partitioning

- A partition is assigned a **list of values**. If the partitioning key has one of these values, the partition is chosen
  - For example all rows where the column Country is either Iceland, Norway, Sweden, Finland or Denmark could build a partition for the Scandinavian countries
- Is also expressed as a simple restriction on the master relation
  - The partitioning predicate involves just one attribute
    - P<sub>I</sub>:City IN ('Hamburg', 'Hannover', 'Berlin')
    - $-P_2$ :City IN (DEFAULT) represents tuples which do not fit to  $P_1$

## Horizontal Partitioning (cont'd.)

#### Hash Partitioning

- The value of a hash function determines membership in a partition. Assuming there are four partitions, the hash function could return a value from 0 to 3
- For each tuple t, of the master-table R, the hash function will associate it to a partition table  $R_i$ 
  - $R_i = \{t_1, ..., t_m/t_j \in R \text{ and } H(t_j) = H(t_k) \text{ for } 1 \le j, k \le m\}$
- This kind of partitioning is particularly used in parallel processing
- The goal is to achieve an equal distribution of the data

## Horizontal Partitioning (cont'd.)

- Horizontal partitioning in DataWarehousing partitions data by
  - Time dimension
    - Periods, such as week or month can be used or the data can be partitioned by the age of the data
    - E.g., if the analysis is usually done on last month's data the table could be partitioned into monthly segments
  - A dimension other than time
    - If queries usually run on a grouping of data: e.g., each branch tends to query on its own data and the dimension structure is not likely to change then partition the table on this dimension
  - On table size
    - If a dimension cannot be used, partition the table by a predefined size. If this method is used, metadata must be created to identify what is contained in each partition

## **Vertical Partitioning**

#### Vertical Partitioning

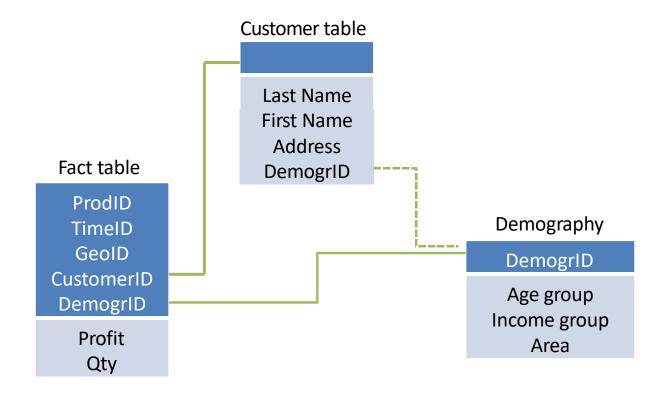
- Involves creating tables with fewer columns and using additional tables to store the remaining columns
- Different physical storage might be used e.g., storing infrequently used or very wide columns on a different device
- Usually called row splitting
- Row splitting creates a one-to-one relationship between the partitions

- In DW, common vertical partitioning means
  - Moving seldom used columns from a highlyused table to another table
  - Creating a view across the two newly created tables restores the original table with a performance penalty
    - However, performance will increase when accessing the highly-used data e.g., for statistical analysis

- In DWs with very large dimension tables (e.g., Amazon - a customer table with tens of millions of records) we have
  - Most of the attributes are rarely, if at all, queried
  - E.g., the address attribute is not as interesting as evaluating customers per age-group
  - But we must still maintain the link between the fact table and the complete customer dimension, which has high performance costs!

- The solution is to use **Mini-Dimensions**, a special case of vertical partitioning
  - Many dimension attributes are used very frequently as browsing constraints
    - In big dimensions these constraints can be hard to find among the lesser used ones
  - Logical groups of often used constraints can be separated into small dimensions which are very well indexed and easily accessible for browsing

• Mini-Dimensions, e.g., the **Demography table** 



- All variables in these mini-dimensions must be presented as distinct classes
- The key to the mini-dimension can be placed as a foreign key in both the fact table and dimension table from which it has been broken off
- Mini-dimensions, as their name suggests, should be kept small and compact

## **Partitioning**

#### Advantages

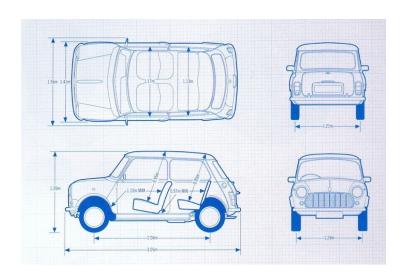
- Records used together are grouped together
- Each partition can be optimized for performance
- Security, recovery
- Partitions stored on different disks:contention
- Take advantage of parallel processing capability

#### Disadvantages

- Slow retrieval across partitions (expensive joins)
- Complexity

- Use partitioning when:
  - A table is > 2GB (from Oracle)
  - A Table is > 100 Million rows (praxis)
  - Think about it, if table is > I million rows

 Partitioning does not come for free!



- Partitioning management
  - Partitioning should be transparent outside the DBMS
    - The applications work with the Master-Table at logical level
    - The conversion to the physical partition tables is performed internally by the DBMS
    - It considers also data consistency as if the data were stored in just one table
  - Partitioning transparency is not yet a standard. Not all DBMS support it!

## **Partitioning Management**

#### Partitions in practice

- Oracle supports Range-, List-, Hash-, Interval-, System-Partitions as well as combinations of these methods
- E.g., partitioning in Oracle:

```
    CREATETABLE SALES(

                                  ORACLE
       ProdID NUMBER,
       GeoID NUMBER,
       TimeID DATE,
       Profit NUMBER)
       PARTITION BY RANGE(timeID)(
              PARTITION before 2008
                     VALUES LESSTHAN (TO DATE ('01-JAN-
                     2008', 'DD-MM-YYYY')),
              PARTITION 2008
                     VALUES LESSTHAN (TO_DATE ('01-JAN-
                     2009', 'DD-MM-YYYY'))
 );
```

## Partitioning Management (cont'd.)

- Partitions in practice
  - In Oracle partitioning is performed with the help of the LESSTHAN function. How can we partition data in the current year?
    - ALTERTABLE Sales

ADD PARTITION after 2008 VALUES LESSTHAN

(MAXVALUE);



# Partitioning Management (cont'd.)

RowID	ProdID	GeoID	TimeID	Profit
121	132	2	05.2007	8K
122	12	2	08.2008	7K
123	15	1	09.2007	5K
124	14	3	01.2009	3K
125	143	2	03.2009	1,5K
126	99	3	05.2007	1K

RowID	ProdID	GeoID	TimeID	Profit
	•••	***	***	
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## **Summary**



- Partitioning: Horizontal or Vertical
  - Records used together are grouped together
  - However: slow retrieval across partitions
  - Mini-Dimensions

## **Next Lecture**

- I. Joins
- 2. MaterializedViews



