

CS 133: Databases

Fall 2019
Lec 26 – 12/12
Data Analytics

Prof. Beth Trushkowsky

Warm-up Exercise

(See exercise sheet. You can start before class.)

Number of reservations for each sid,bid pair

Goals for Today

- Understand how Analytics processing (OLAP) is different than Transactional processing (OLTP)
- Reason about how data is organized and queried in a data warehouse
- Discuss current trends in Big Data processing

Data Analytics and Decision Support

- **Idea:** current and historical data to identify useful patterns and support business strategies
- Complex, interactive, exploratory analysis of data
 - Large datasets
 - Data integrated from across all parts of an enterprise
 - Data is fairly static
- **OLAP:** on-line analytical processing
 - In contrast to **OLTP** (on-line transactional processing)

OLAP vs. OLTP

- **OLTP**
 - Update-heavy
 - Short, simple transactions
 - Goal: transaction throughput
- **OLAP**
 - Mostly reads
 - Longer, complex queries for analysis and decision-making
 - Goal: fast queries

Data Integration

- Data may reside in many distributed, heterogeneous OLTP sources
 - Sales, inventory, customer, ...
 - NC branch, NY branch, CA branch, ...
- Need to support OLAP over integrated view of the data
- Possible approaches to integration
 - *Eager*: integrate in advance and store the integrated data in a **data warehouse**
 - Need ETL
 - *Lazy*: integrate on demand; process queries over distributed sources—the approach of mediated or federated systems

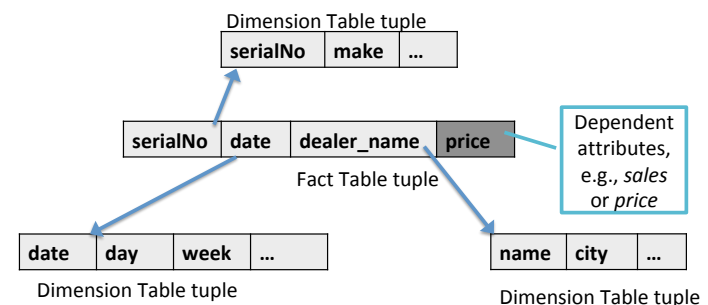
Example: Car Sales Schema

```

Cars(serialNo, make, model, color)
Dealers(name, city, state, phone)
Date(date, day, week, quarter, month, year)
Sales(serialNo, date, dealer, price)
    
```

Star Schema in Relational OLAP (ROLAP) System

- Fact table **BCNF**; dimension tables possibly **denormalized**
 - Dimension tables are small; updates/inserts/deletes are rare... anomalies less important than performance
- **Star Schema**



A Multidimensional View

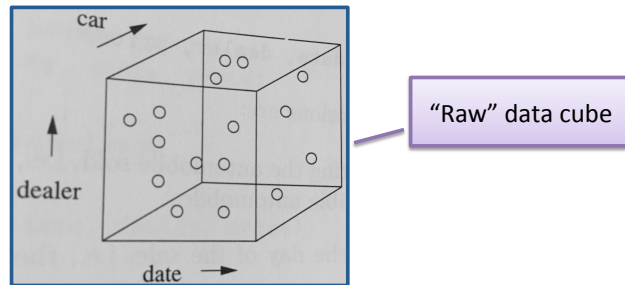
- Example car sales schema:

```
Cars(serialNo, model, color)
Dealers(name, city, state, phone)
Date(date, day, week, month, year)
```

Conceptual
dimension
table

```
Sales(serialNo, date, dealer, price)
```

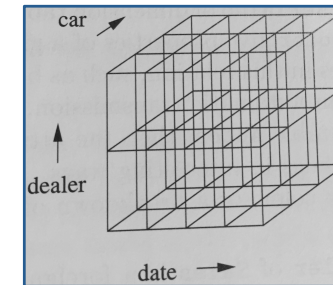
Fact table



Dicing the Cube

Think SQL "group by"

- Can think of partitioning the raw data cube along each dimension at some level of granularity
- A choice of partition for each dimension "dices" the cube

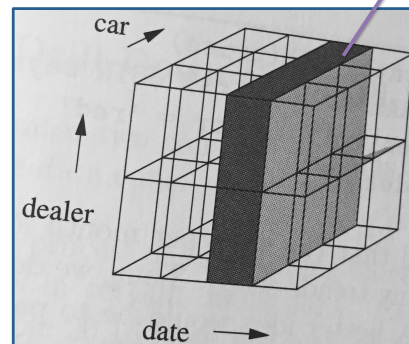


Slicing the Cube

- Idea: want info about a fixed *slice* of the data

Sales data for
one date slice
(e.g., one year)

- In general, in SQL:
 - **Dice**: GROUP BY
 - **Slice**: WHERE



Example: Data Analysis

```
Cars(serialNo, make, model, color)
Dealers (name, city, state, phone)
Days(date, day, week, quarter, month, year)
Sales(serialNo, date, dealer_name, price)
```

- Suppose *Mazda3* model is not selling as well as anticipated
- **Query**: which colors not doing well?

```
SELECT color, SUM(price)
FROM Sales NATURAL JOIN Cars
WHERE model = "Mazda3"
GROUP BY color;
```

Exercise 2 (a-c)

(a)
 SELECT color, month, SUM(price)
 FROM Sales, Cars, Days
 WHERE Sales.serialNo = Cars.serialNo
 AND Sales.date=Days.date
 AND model = "Mazda3"
 GROUP BY color, month;

(b)
 SELECT dealer_name, month, SUM(price)
 FROM Sales, Cars, Days
 WHERE Sales.serialNo = Cars.serialNo
 AND Sales.date=Days.date
 AND model = "Mazda3"
 AND color = "red"
 GROUP BY month, dealer_name;

(c)
 SELECT dealer_name, year, SUM(price)
 FROM Sales, Cars, Days
 WHERE Sales.serialNo = Cars.serialNo
 AND Sales.date=Days.date
 AND model = "Mazda3"
 AND color = "red"
 AND (year = 2016 OR year = 2017)
 GROUP BY year, dealer_name;

Analysis: Cross-tabulation

Also called
"pivoting"

- Sales from each dealer by car color
 – View popularized by spreadsheet applications

		Car color			
		Red	White	Blue	<i>total</i>
Dealer	Alice	90K	30K	120K	240K
	Bob	100K	10K	40K	150K
	<i>total</i>	190K	40K	160K	390K

How many SQL queries to generate the data in this table?

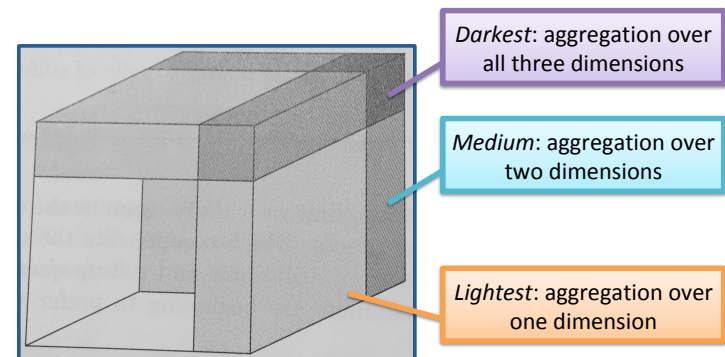
OLAP Queries

- A common operation is to **aggregate a measure** over one or more dimensions.
- **Roll-up**: Aggregating at coarser granularity, e.g., higher level in dimension hierarchy.
- **Drill-down**: The inverse of roll-up

Specialized *MOLAP* and *ROLAP* systems may store pre-aggregated data (materialized views)

The Data CUBE Multidimensional OLAP (MOLAP)

- A **CUBE** relation:
 generalization of the cross-tabulation



Analyzing Big Data: Current Trends

- Motivation
 - Expensive ROLAP and MOLAP systems not for everyone
 - Desire to analyze semi-structured or unstructured data
- **Big Data** rampant!
 - E.g., data sets generated by some of the applications backed by NoSQL systems
 - Sensor data, tweets, etc.
- **Trend:** many people using **MapReduce/Hadoop** for Big Data Analysis
 - Scalability and commodity hardware

Open-source version of
Google's MapReduce

Pokémon or Big Data?

<https://pixelastic.github.io/pokemonorbigdata/>



The screenshot shows a web browser window with the URL <https://pixelastic.github.io/pokemonorbigdata/>. The page content includes the text "Hadoop is Big Data!" in large blue letters, a yellow cartoon elephant, and the text "Hadoop is a distributed system for counting words." Below this is a blue button labeled "Next question".

Final Exam: Logistics

- Take-home exam
- Due to my office (Olin 1267) at or before Wednesday, December 18th, 5:15pm
- Two 8.5x11, double-sided note sheets
 - You can use your note sheet from the midterm as one of the two
 - No other resources
- 3-hour timed exam

Possible Topics on Final

- Cumulative-ish
 - Topics we covered earlier still relevant (e.g., hash & tree indexes, estimating cost in I/Os)
 - Won't focus on nitty gritty from before midterm (e.g., linear vs extendible hashing)
- Query Optimization
- Transactions and ACID
- Database design
- ORDBMS, Distributed DBMS and NoSQL, OLAP (high-level)
- General themes
 - Reasoning about cost and tradeoffs
 - Consistency and correctness with concurrent access and failures

Query Optimization

- Query
 - relational algebra tree
 - logical plan
 - physical plan
- Unit of optimization: query block
- Logical plan
 - Relational algebra equivalences
 - Outer vs. inner relation in joins
 - Query plan tree shape: bushy, linear, deep

Query Optimization

- Choosing physical plan
 - Enumerate plan space
 - Join permutations and orders
 - System R choices
 - Estimate cost of plan
 - Picking cheapest
 - Dynamic programming algorithm (idea)
 - Interesting orders
- Cost estimation
 - Operator algorithm cost
 - Estimating cost of different join algorithms
 - Operator result size estimation
 - Selectivity/Reduction Factor, statistics, histograms
 - Using indexes

ACID Transactions

- Transactions, how to achieve ACID
- Isolation (I)
 - Schedules: serializable, conflict-serializable, etc.
 - Anomalies from interleaved actions, conflicting actions
 - Locking, lock granularity and compatibility, deadlock detection and prevention
 - 2PL vs Strict 2PL, cascading aborts
 - Optimistic concurrency control, backwards validation algorithm
- Recovery (A and D)
 - Steal vs. force and implications on UNDO/REDO
 - Write-Ahead-Logging
 - ARIES recovery algorithm

Database design

- E/R modeling (general idea)
 - Entities, relationships, weak entities
 - Capturing key and participation constraints
- Functional dependencies
 - Attribute closure, Armstrong's axioms
 - Determining candidate keys
 - Role in detecting data redundancy
- Schema refinement
 - Normalization
 - BCNF normalization process
- Capturing integrity constraints in relational schema
- General motivation and ideas from ORDBMS

Special Topics

- Distributed DBMS
 - Goals of data partitioning and data replication
 - Types of partitioning: range vs hash
 - Replication
 - Synchronous vs asynchronous
 - Strong vs. eventual/weak consistency
 - Challenges with distributed xacts (generally)
- NoSQL
 - CAP theorem
 - Query restrictions for performance (generally)
- Analytics
 - Generally what OLAP is, vs. OLTP, and what kinds of queries run