Data-Management for Data-intensive Computing

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CMPSC 274: Advance Topics on Database System

Divyakant Agrawal Department of Computer Science University of California at Santa Barbara

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Prelude

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- Data management is at the cross-roads:
 New models for data-intensive computing
 Significant turmoil in terms of technological advances
 Rapid changes have presented the data management research community with unprecedented challenges.
- Inevitable to re-examine the context in which data management evolved in **the past**.
- In the same vein, we need to explore the role of data management in **the future**.
- Course Objective: Comprehensive understanding of data management and data analysis paradigms.

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Data-intensive Computing

- Storage and retrieval management of persistent data.
- Large-scale data analysis for data-centric decision making.







- First half of the course (assignment/ assessment based):
- Home-works
- Mid-term Exams
- Text book and Other Refereces
- Second half of the course:
- Project based
- Large programming/implementation project (2 person)

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÷ **Historical Perspective**

- Advent of computer technology:
- Persistent storage of data and information Value of data/information realized very early especially in the context of business entities

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- Early efforts in the industry:
- Effective data management solutions
 Based on storing data:
 Files: logical abstraction
 Tapes: physical realization
- File based data management
- Problems in accessing data
- Problems in processing dataIn general, problems in effective usage of data

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Historical Perspective

- Emergence of alternative storage models:
- Departure from file-based storage
 New data models to enable data access based on its
- attributes
- New language models to enable effective manipulation of data
- Data models (circa 1965):
- Network model: essentially to model business entities using the information paradigm Hierarchical model: another variant
- Standardization efforts:
- Economies-of-scale/minimize duplication
- CODASYL





÷ Codd'69: Relational Data Model

- Relational Data Model:
- Tabular framework for data representation
- Intuitive and easy to comprehend
 Complete theoretical framework
- Relational algebra (operational framework):
- An algebraic framework for operating on relational data
- Well-defined algebraic operators
- Relational calculus (theoretical framework):

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- First-order logic
- Declarative querying framework
- Equivalent to relational algebra

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Relational Data Model

- RDBMS model became extremely successful.
- Logical Data Model:
- Intuitive
- Well-defined
- Design time considerations need not focus on physical issues
- Physical storage independence:
- Run-time system maintains the access methods
- Dynamic mapping from logical to physical level
- Declarative Query Interface:
- Users did not need to be expert programmers

Data Management Evolution

- RDBMS became highly successful:
 Widely adopted by both large and small business entities
- Enterprises became increasingly reliant on databases
- Primarily used for day-to-day operations:

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- Banking operations
- Retail operations
- Travel industry

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Batched Transaction Processing

- Nomenclature:
- Transaction Processing Systems
- Typical usage:
- Spool client transactions during the day
- During the night, spooled transactions applied to the database state of previous night
- New database state becomes available for the next working day
- Advantages:
- Almost up-to-date information on the finger-tips
- Failure-recovery is in-built in the paradigm
- No issues of concurrency

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÷ Concurrency & Failures: A Quick Preview

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- List maintenance:
- lookup/find operations in O(log N) time.
- Read-only operations: Concurrency does not cause any difficulties.
- List updates:
- Inserts/deletes also in O(log N) time if executed sequentially.
- What if I specify that operations are arbitrarily
- interleaved? Worse yet: what happens if the updaters can fail?
- Can you do it safely? Do you have the necessary tools to solve this problem?

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Data Analysis (90s) → Business Intelligence (00s) → Big Data (Now!!!)



÷ 50 Years of Business Intelligence

- Vision of Business Intelligence:
- Hans Peter Luhn in a 1958 article.
- Predates the notions of Databases and Data Management.
- A pioneer in Information Sciences:
- New use of the term *thesaurus*
- Automatic creation of literature abstracts
- 16 digit Luhn's number widely used for credit
- cards and other banking instruments

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4 Luhn's Vision Defined BI as:

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"... provides means for selective dissemination to each of its action points in accordance with their current requirements or desires."

- Key technologies:
- Auto-abstracting of documents,
- Auto-encoding of documents, and Auto creation and updating of profiles

Breadth of the vision:

"... business is a collection of activities carried on ... be it science, technology, commerce, industry, law, government, defense, et cetra."

"... intelligence is also defined ... as the ability to apprehend the interrelationships of presented facts in such a way as to guide action towards a desired goal."

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The Early Years (1970s-1980s)

- Contrary to Luhn's overarching vision early efforts on business information remained focused on *database management technology*.
- With the advent of the relational model:
- DBMS technology became pervasive and matured.
- Widely adapted by most enterprises.
- Online Transaction Processing became a proven paradigm for business operations.
- Consequence:
- Massive proliferation of OLTP systems especially within a single enterprise.
 Data-driven decision making became a norm.
- Disparate reporting from multiple operational data sources.

÷ Notion of "Data Analytics" (1990s)

Presence of multiple operational systems created a *fractured* view of an enterprise.

Devlin & Murphy introduced the term business data warehouse in 1988:

- A unified view of the enterprise primarily for integrated reporting.
- Catalysts:
- Demand for reporting key factors being PCs and spread-sheets. Market potential - Teradata, Red-brick Systems, etc.
- Negative factors:
- Unproven, immature, and expensive technology proposition.
 Distinction between DBMS and DW: no clarity, *?duplication?*
- Fairly laborious and time-consuming data integration process
 No clear stake-holders
 ^{2nd} Class Entity often resulting in adversarial atmosphere.
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Data Warehousing: Current State

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Keys to success:

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- Enormous contribution of DW evangelist Ralph Kimball
- STAR schema & Dimensional model for DW: intuitive and scalable No compromise on the autonomy of operational data sources
- Persisting head-winds:
- Since does not directly contribute to P&L:
- ROI question still persists.
- Not a plug & play technology:
 Very high consulting costs.
- Legacy of significant time and cost over-runs of most data warehousing projects.
- Batch-oriented DW Architecture:
- Deemed too costly just for integrated reporting.Needed intuitive analytical capabilities.

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÷ Hither "Business Intelligence" (2000-)

- Gray et al. [1996] introduced the CUBE operator for roll-up and drill-down analysis of multi-dimensional data (i.e., DW Model).
- DW enterprises (Hyperion, Cognos, Analysis Services, etc.) adapted the CUBE architecture and called it:
- business intelligence.
- Problem:
- Early BI (CUBE) technology had serious issues of scaling → only accentuated the ill-repute of DW/BI technologies
 Underlying problem: exponential explosion of data storage

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Hence, perhaps the term Real-time Business Intelligence

 to convey the "criticality" of such technology to business
 leaders.

Current debate: what exactly is meant by "real-time" in Business Intelligence?
 In 2006, in this workshop, Donovan Schneider – gave numerous examples of "degree of timeliness" for a variety of analysis tasks.
 My personal view is that the correct term should have been: Online Business Intelligence.

Assuming that – redefine the DW/BI architecture to support RTBI.

Concluding Remarks Data Management:



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- Will study the models, paradigms, theory, and algorithms needed for Enterprise Scale Data Management (& application development)
- Will then examine the disruption that has occurred with the Internet and Web-based application:
- underlying factors for this disruption and
- Proposed solution

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