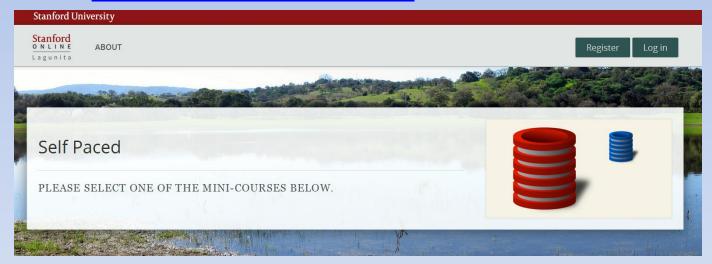
Remember: Stanford Online

https://lagunita.stanford.edu/courses/DB/201
 4/SelfPaced/about



engineering.stanford.edu/about/dean

Dean

Jennifer Widom, Frederick Emmons Terman Dean of the School of Engineering

Jennifer Widom, a professor of computer science and of electrical engineering at Stanford University for more than two decades, became dean of the School of Engineering in March 2017.



cs.stanford.edu/people/widom/



Jennifer Widom

Frederick Emmons Terman Dean of the School of Engineering Fletcher Jones Professor in Computer Science and Electrical Engineering

Stanford University

- in March 2017 I became Stanford's 10th Dean of Engineering. News release here and new home page here
- I spent my 2016-17 fall-winter sabbatical traveling the world offering free short-courses and workshops. Details here
- For an update on Stanford's MOOC offering in Databases, please visit this page

Update on Stanford's Databases MOOC

Jennifer Widom

Current Status

Stanford's free online offering in Databases is now available as a set of self-paced "mini-courses" created from the original *Introduction to Databases* course (see <u>History</u> below). All of the mini-courses are hosted on the <u>OpenEdX platform</u>, with a starting page <u>here</u>.

Recognizing that different students have different goals in learning about databases, and that many of the topics are modular and independent, in the spring of 2014 we took the original ten-week *Introduction to Databases* course and broke it into 14 mini-courses. All of the mini-courses are based around video lectures and/or video demos. Many of them include in-video quizzes, stand-alone quizzes, and/or automatically-checked interactive programming exercises. All of them are self-paced, and if a sufficiently high score is achieved, a Statement of Accomplishment is issued.

Fourteen Mini-Courses - by Area

Data Models

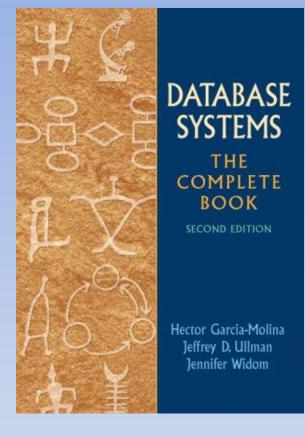
- · Introduction and Relational Databases
- XML Data
- JSON Data

Querying Relational Databases

- Relational Algebra
- SQL

Querying XML Databases

- XPath and XQuery
- XSLT



Database Systems

Tonight:

- Chapter 1: Database World
 - Overview of DBMS

Overview of Database Management System

- Look at System Components
- In Memory Data Structures
- Control/Data Flow
- Data Flow Only

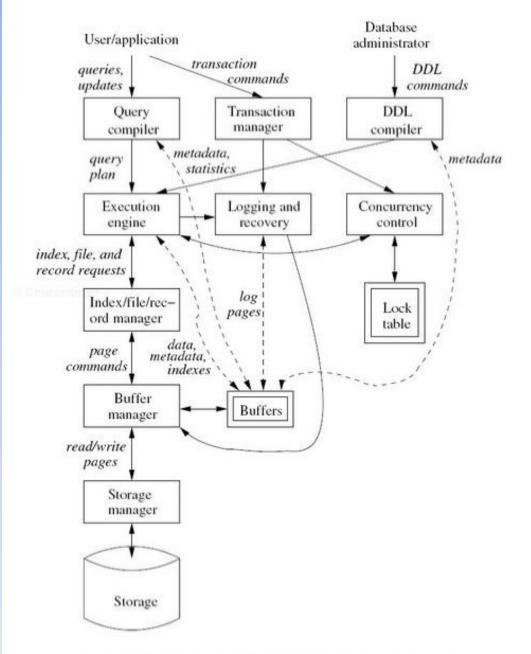
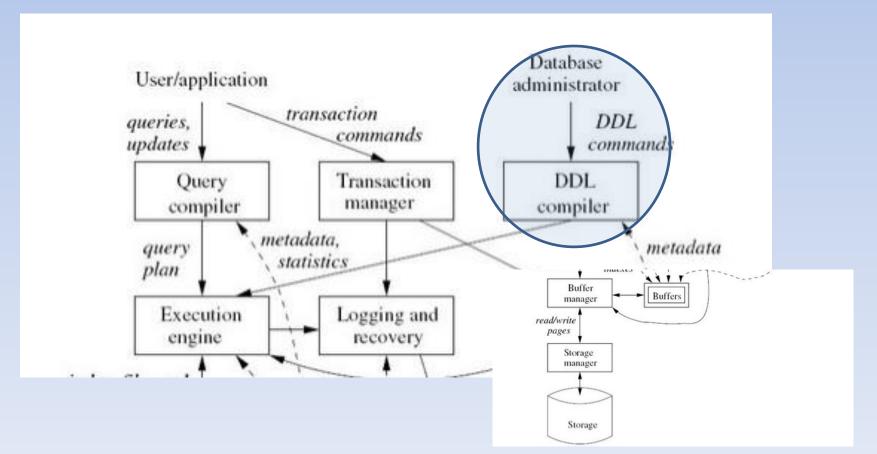


Figure 1.1: Database management system components

Data-Definition Language Commands

 These are the commands that create/change the database structure.

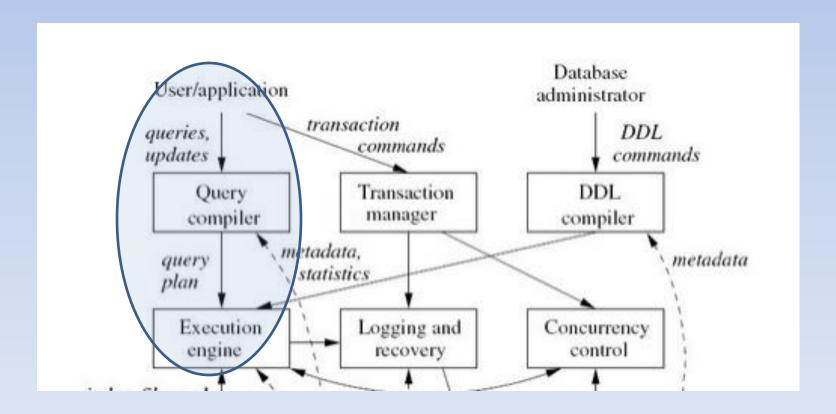


Query Processing w/ Data Manipulation Language

Vast majority of work for DB.

Query Results

- Query Compiler generates Query Plan
- Execution Engine takes Query Plan



Query Compiler w/ Chapter 16

Chapter 16

The Query Compiler

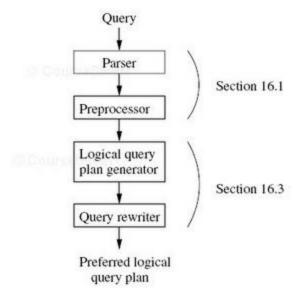


Figure 16.1: From a query to a logical query plan

up the architecture of the query compiler and its optimizer. 5. 15.2, there are three broad steps that the query processor

Query Compiler

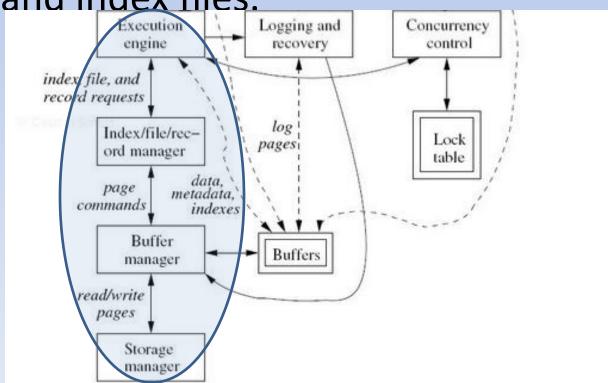
- Query Parser
 - Builds a tree structure from textual query
- Query Preprocessor
 - Does some checking
 - Converts parse tree into Relational Algebra Expression
 Tree
- Query Optimizer
 - Uses Optimization over Expression tree to generate final sequence of optimized operations on the actual data!

Query Results w/ Execution Engine

 Issues request for small pieces of information to resource manager.

Resource manager knows about data files

(tuples) and index files.



Query Execution

Chapter 15

15.2.1 One-Pass Algorithms for Tuple-at-a-Time Operations

The tuple-at-a-time operations $\sigma(R)$ and $\pi(R)$ have obvious algorithms, regardless of whether the relation fits in main memory. We read the blocks of R one at a time into an input buffer, perform the operation on each tuple, and move the selected tuples or the projected tuples to the output buffer, as suggested by Fig. 15.5. Since the output buffer may be an input buffer of some other operator, or may be sending data to a user or application, we do not count the output buffer as needed space. Thus, we require only that $M \geq 1$ for the input buffer, regardless of B.

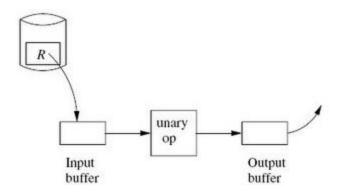


Figure 15.5: A selection or projection being performed on a relation R

Query Execution

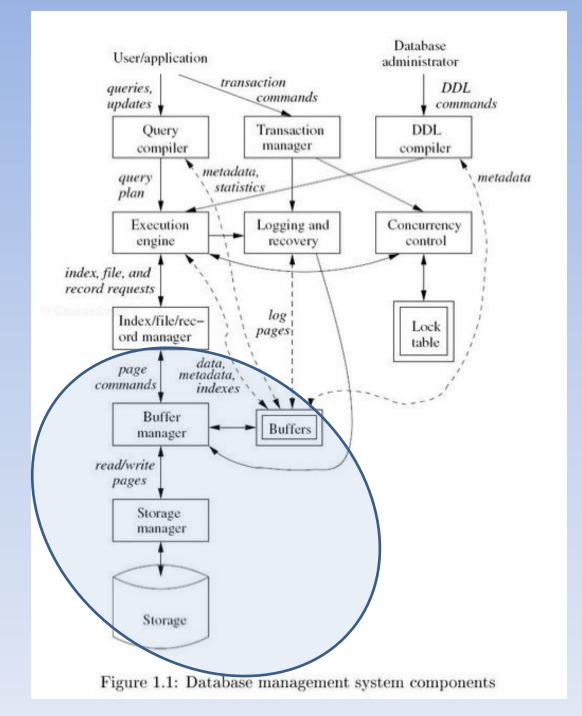
Transaction Processing

- Queries and DML Operations are grouped into Transactions
- These group of operations must be executed
 Atomically and in Isolation
- Actions executed must also be **D**urable
- In addition Consistency constraints must be maintained.
- ACID Properties of Transactions

Storage Manager

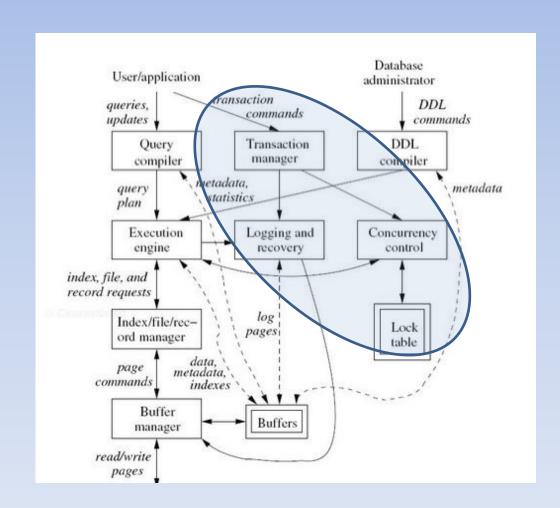
- Storage manager takes care of getting data off of secondary storage and into/out of memory from secondary storage!
- Buffer Manager
 - Partitions memory into buffers
 - Transfers data between disk blocks and memory buffers

- Data:
 - Student Data
 - Movie Ratings
- Metadata:
 - Schema
 - Users
- Log Records
- Statistics
- Indexes



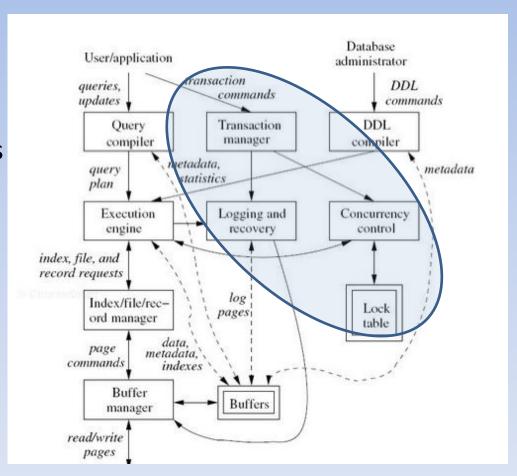
Transaction Processing

- TransactionManager:
 - Atomicity
 - Isolation
- Logging and Recovery
 - Durability



Transaction Processing

- Logging
 - Supports durability
 - Recovery from failures.
 - Rollback from errors.
 - Recovery Manager uses logs to fix db.
- Concurrency Control
 - Insures Isolation
 - Uses Lock table
- Deadlock Resolution
 - Certain configuration create a situation where the system enters "Deadlock"
 - Recognizing deadlock, and resolving it.



Database Systems Study

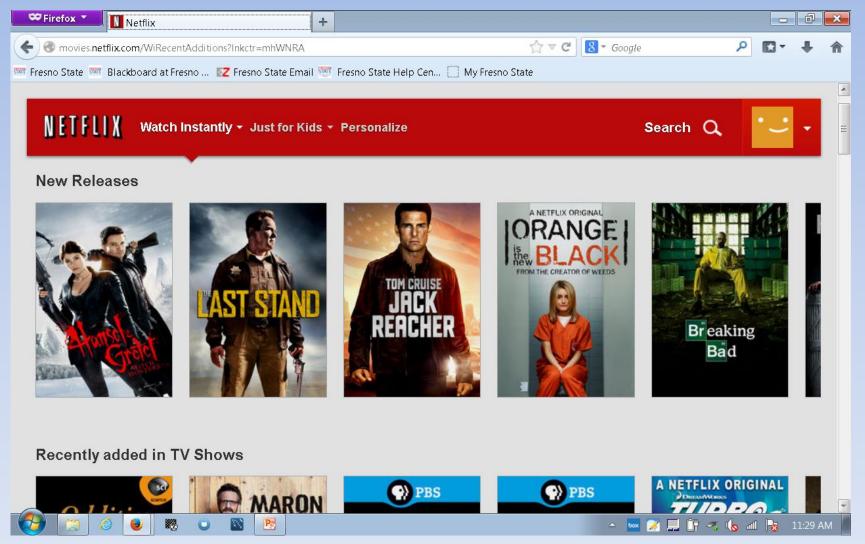
- Part III: Semi-structured Data
 - CH 11, 12
 - XML
 - JSON
- Part II: Database Programming
 - CH 5, 6, 7, 8, 9, 10
 - Relational Algebra
 - Datalog
 - SQL
- Part I: Relational Database Modeling
 - CH 2, 3, 4
 - Relational Data Model
 - Design Theory
 - E/R, UML, ODL

Database Systems Study

- Part IV: Database Systems Implementation
 - CH 13, 14, 15, 16, 17, 18, 19, 20
 - Secondary Storage
 - Index Structures
 - Query Execution, Query Compiler
 - System Failures, Concurrency Control, Transaction Management
 - Parallel and Distributed Databases
- Part V: Other Issues
 - CH 21, 22, 23
 - Information Integration
 - Data Mining
 - Database Systems and Internet

Some of My Favorite DB's

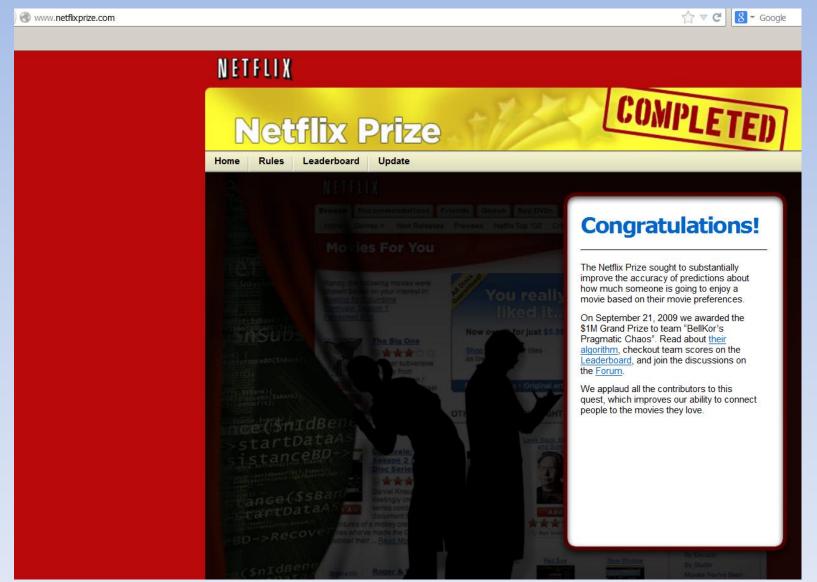
Database Example: Netflix



Netflix

- Over 40 million subscribers
- More than 75000 titles
- IMDB has 2.7 million movies
- IMDB has 5.6 million movie related personalities

Netflix / Data Mining

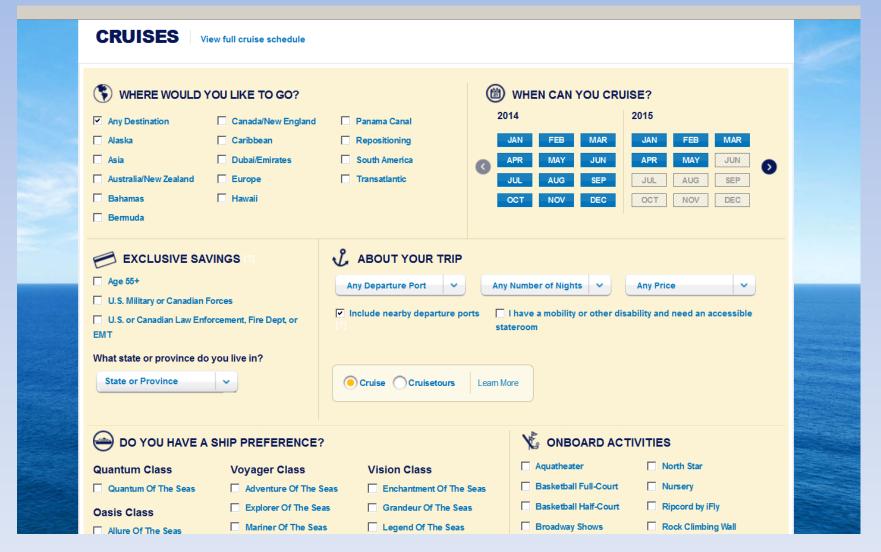


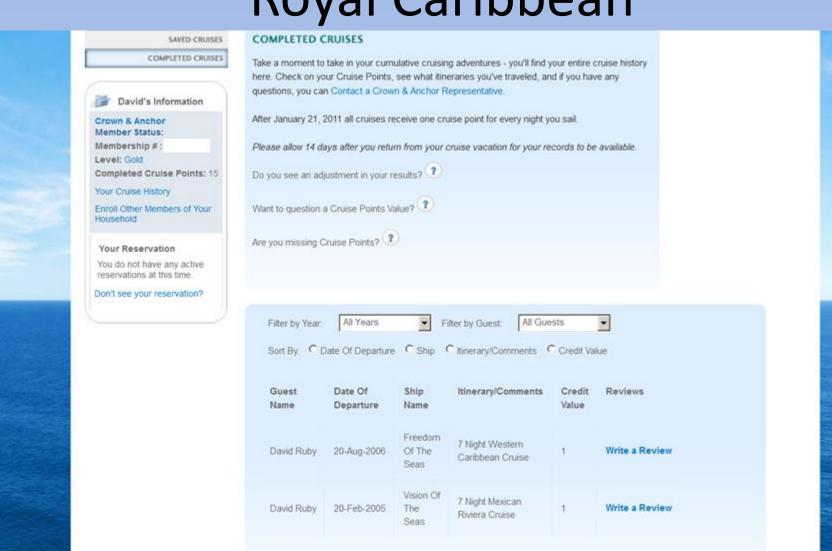
Data Mining: Netflix Data Movie Recommendations: Netflix Challenge (excerpts)

- We're quite curious, really. To the tune of one million dollars.
- We've developed our world-class movie recommendation system: Cinematch. Its job is to predict whether someone will enjoy a movie based on how much they liked or disliked other movies.
- We use those predictions to make personal movie recommendations based on each customer's unique tastes.
- And while Cinematch is doing pretty well, it can always be made better.

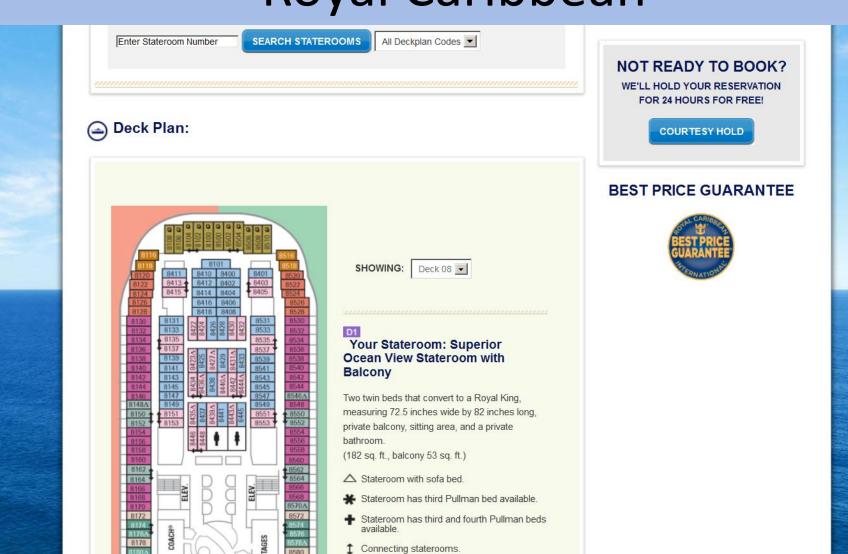
Reservation Systems

- Required Data:
 - Browsing
 - Reserving
- Other opportunities w/ Data





2 Results



Indicator accoscible statorooms

David's Information

Crown & Anchor

Member Status:

Membership #:

Your Cruise History

Your Reservation

Enroll Other Members of Your

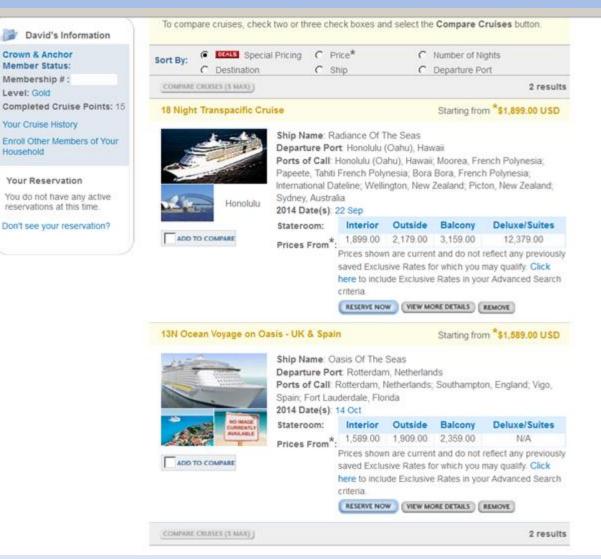
You do not have any active

Don't see your reservation?

reservations at this time.

Level: Gold

Household



Royal Caribbean / Data Mining

- Which cruises were most popular last year?
- Given trends for each major cruise destination over the last 10 years, which destination should be most popular next year.
- Given the list of cruises for an individual, what is the predicted probability for their next cruise choice.
- Others????

Database Examples

- Credit Reports (Experian/TransUnion/Equifax)
- Facial Recognition (Face Print) Databases
 - Facebook, Picassa, etc...
- FBI 12.8 Million Mugshots in new facial recognition database.
- NSA Database of cell phone calls
- New National Healthcare Website Database

NSA Database of Phone Logs

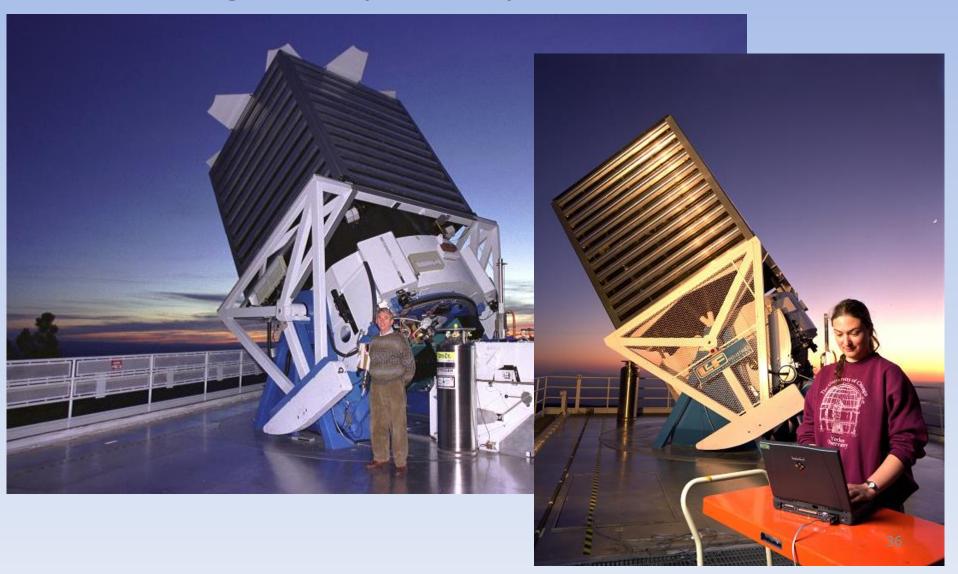
- Data Warehousing:
 - Nightly dumps of phone logs from each phone carrier in US to central warehouse
- Phone Carriers all registered

NSA Database / Data Mining

- What area in America is most likely to place a call to Switzerland?
- Given a list of calls for an individual, what is the probability they will make a call to Switzerland.
- Others????

eScience

Sloan Digital Sky Survey:



eScience

- Sloan Digital Sky Survey:
- Dedicated 2.5-meter telescope
- A pair of spectrographs fed by optical fibers measured spectra (hence distances) of more than 600 galaxies and quasars in a single observation
- 80TB (80 * 10^12 bytes) raw image data over a 7 year period

SDSS Database

- About the Database
- http://skyserver.sdss3.org/dr8/en/help/docs/intro.asp#sqlcl
- Commercial Relational Database Management system (DBMS) Microsoft's SQL Server.
- SDSS Command Line Query Tool

Author: Tamas Budavari, JHU

Date: April 2003, updated January 2005 (DR3) **sqlcl.py** is a (very) simple <u>Python</u> program that can run SQL queries against <u>SkyServer</u>. Python runs on your favourite OS including the most exotic ones. The query goes through the same .asp page that you use in the web form or using *wget*.

• http://skyserver.sdss3.org/dr8/en/help/download/sqlcl/default.asp

SDSS sqlcl.py

```
formats = ['csv','xml','html']
astro url='http://skyserver.sdss3.org/dr8/en/tools/search/x sql.asp'
public url='http://skyserver.sdss3.org/dr8/en/tools/search/x sql.asp'
default url=public url
default fmt='csv'
def query(sql,url=default_url,fmt=default_fmt):
  "Run query and return file object"
  import urllib
  fsql = filtercomment(sql)
  params = urllib.urlencode({'cmd': fsql, 'format': fmt})
  return urllib.urlopen(url+'?%s' % params)
```

Database Roles

- Database Administrator
- Database Systems Programmer

DB2 Systems Programmer

Database Analyst olnshire, US-IL (Greater Chicago Area)





- Database Applications Developer
 - Aon Hewitt is the global leader in human capital consulting and outsourcing solutions. The company
- Data Analyst

DB2 Analyst with Provisioning experience

Advance Team Sicesering Norteapolis, MN (Greater Minneapolis-St.



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As one of the world's leading staffing companies, our primary objective is clear-cut: To perfectly align the best people with appropriate positions for our clients everywhere. Every day, we talk to hiring managers, HR directors, procurement managers, and senior management to learn their needs. At the same time, we're bringing in candidates and associates with a variety of skills and backgrounds, seeing how we can best put their skills to use.

Check out below this exciting DB2 Provisioning Analyst opportunity with our direct client for their Minneapolis, Minnesota location!

Ided in enterprise services, boast expert lobal setting, possess good ly and in team settings. You also provide ing IT systems, applications, processes ses and standards. You will be part of a th Oracle 10G/11G and MS SQL Server allation, troubleshooting, performance

Key Features of the Database

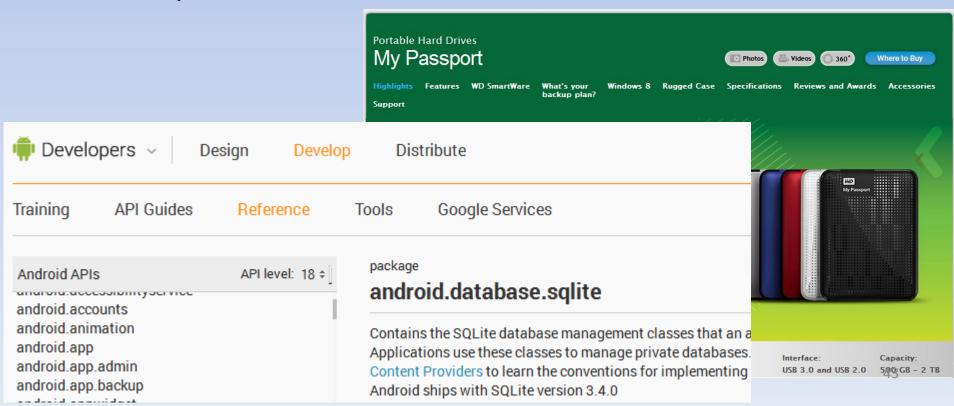
- Store large amounts of data (terabytes)
 - -1 Kilobyte = 1,000 bytes
 - 1 Megabyte = 1,000,000 bytes
 - -1 Gigabyte = 1,000 * 1 MB = 1,000,000,000 (10^9)
 - 1 Terabyte = 1000 * 1GB = 1, 000, 000, 000, 000 (10^12)
 - 1 Petabyte = 1000 *1TB = 1, 000,000, 000,000, 000 (10^15)
- Durability
 - Logging
- Concurrent Users
 - Transactions

Database History

- Early History: File System Hierarchical
- 1960's First DB's (Banking, Reservations, Corporate Records)
 - Extension of File System
 - Hierarchical
 - Programmatic Difficulty
- 1970: TED CODD gave the world Relations
- Relations had powerful mathematical properties
- Relations provided Data Abstraction
- 1990's RDMS are The Norm

Database History

- Smaller & Smaller:
 - MySQL for multiple PC Platforms
 - SQLite for Android
 - Terabyte Drives for PC's



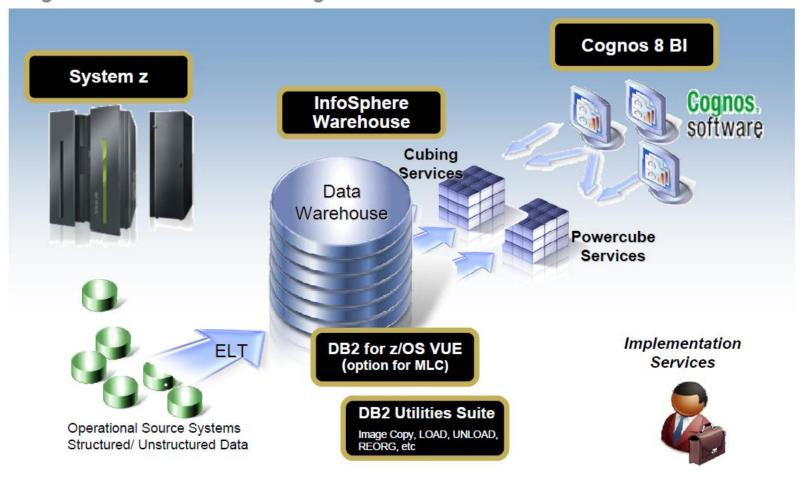
Database Future

- Bigger & Bigger
 - Google holds petabytes of data from web crawls
- Information Integration
 - Legacy Systems
 - Data Warehousing
- Parallelism:
 - Hadoop

Data Warehousing

IBM Smart Analytics System 9600

High Value Data Warehousing

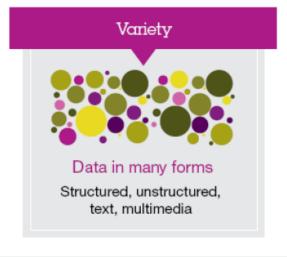


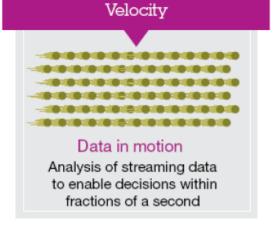
17

Big Data

Characteristics of big data

Data at scale Terabytes to petabytes of data





Veracity

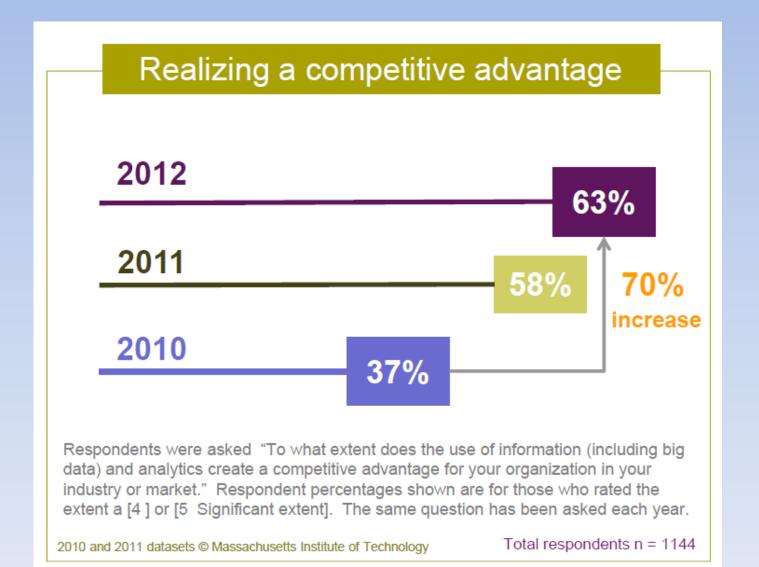


Data uncertainty

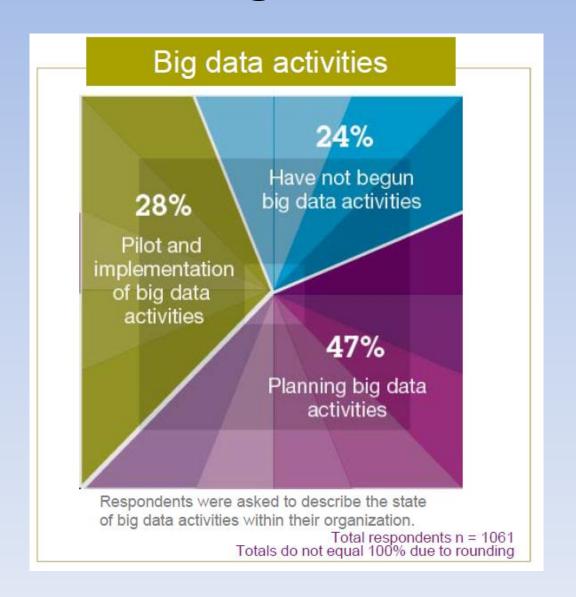
Managing the reliability and predictability of inherently imprecise data types



Big Data



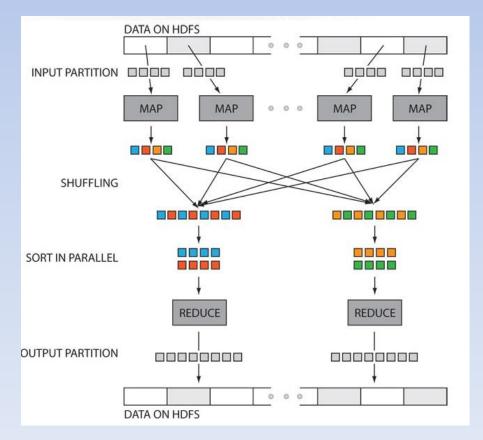
Big Data



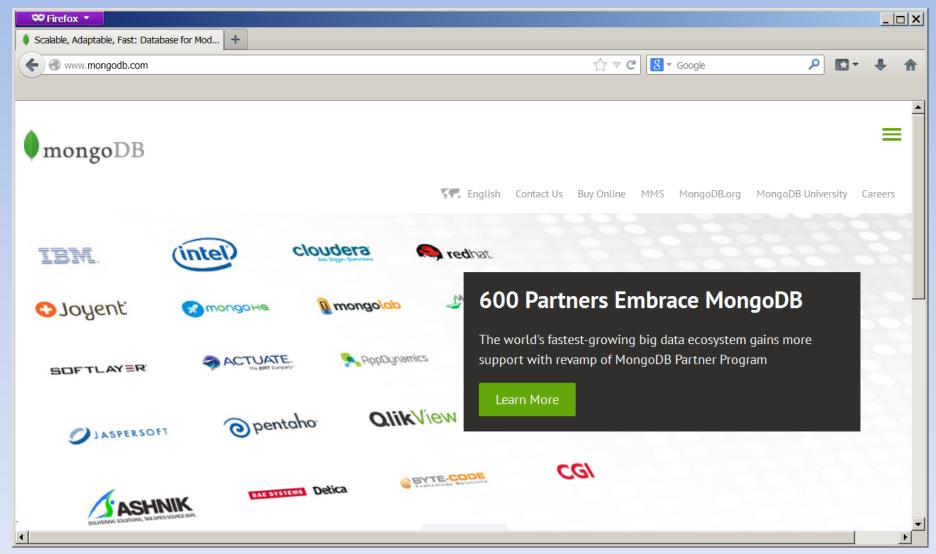
Parallelism: Hadoop

Hadoop is both a parallel computing system and a parallel

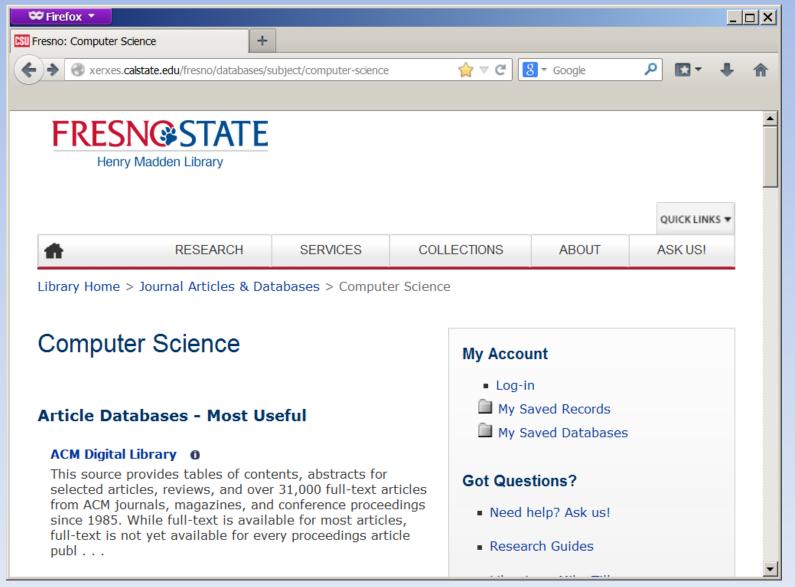
database.

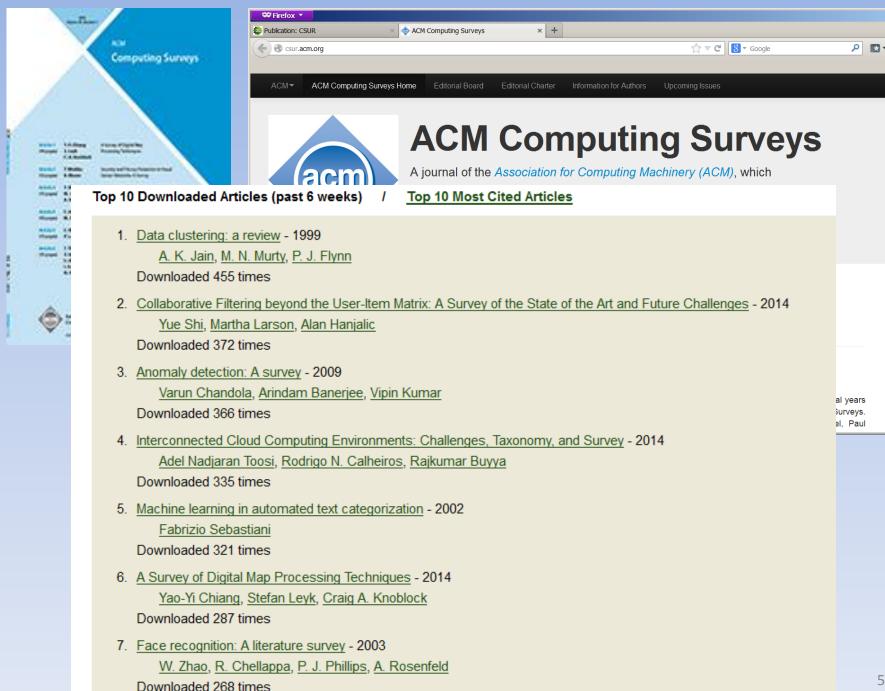


MongoDB



Research





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 Ronald Fagin

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Author: E. F. Codd IBM Research Lab, San Jose, CA

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Pages 397-434

ACM New York, NY, USA

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Publication:



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Current Issue

Volume 8 Issue 3 (June 2014)

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Efficient Discovery of the Most Interesting Associations

Full Text: 🔁 PDF

Authors: Geoffrey I. Webb Monash University, Australia

Jilles Vreeken University of Antwerp, Belgium

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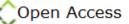
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2013 Article

- Research
- Refereed





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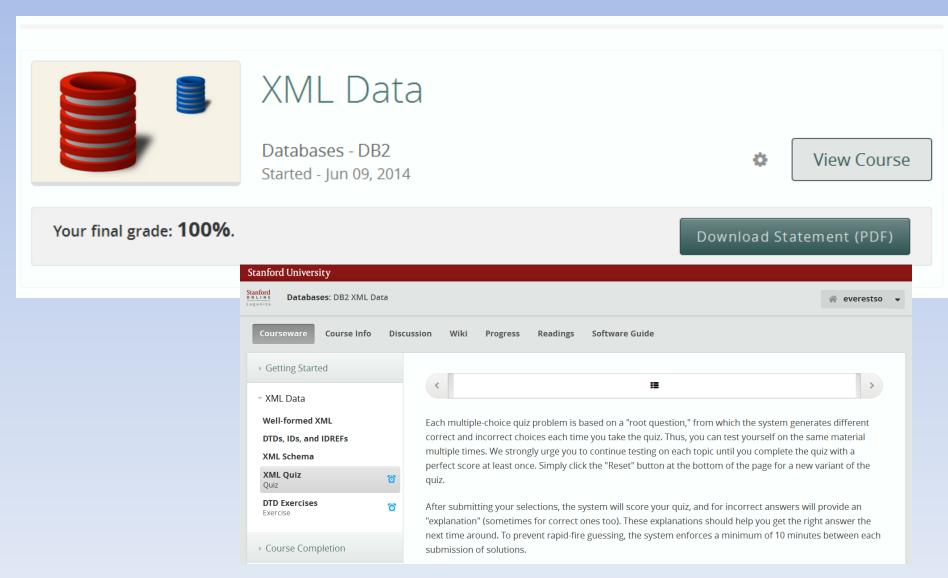
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