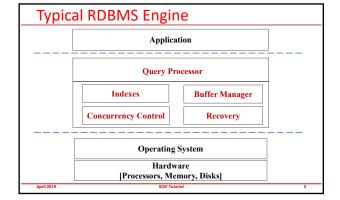


Relational DBMS

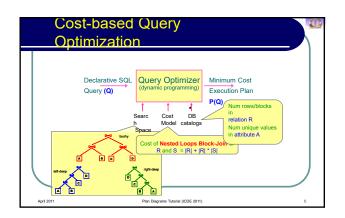
- · Workhorse of today's Information Industry
 - Commercial
 - IBM DB2, MS SQL Server, Oracle Exadata, HP SQL/MX
 - Public-domain
 - · PostgreSQL, MySQL, Berkeley DB
- · Extensively researched for over four decades

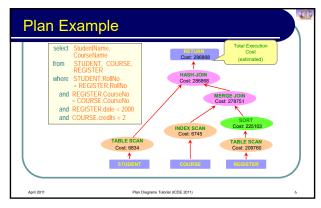
 - JournalsACM TODS, IEEE TKDE, VLDBJ, ...
 - Conferences
 - ACM SIGMOD, IEEE ICDE, VLDB, EDBT, CIKM, ...

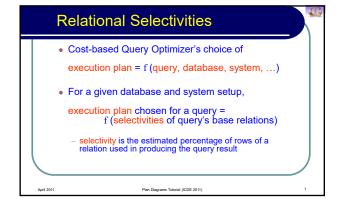


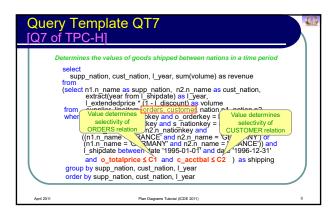
Design of RDBMS Engines

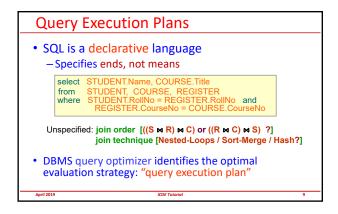
- Transaction Processing (ACID)
 - WAL/ARIES for Atomicity/Recovery
 - 2PL for Concurrency Control
- · Data Access Methods
 - B-trees/Hashing for Large Ordered Domains
 - Bitmaps for Small Categorical Domains
 - R-trees for Geometric Domains
- Memory Management
 - LRU-k (k=2 balances history and responsiveness)
- Query Processing (SQL)
- –"Black Art"

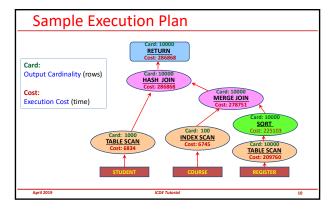


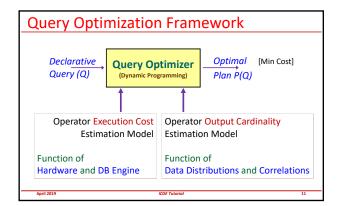




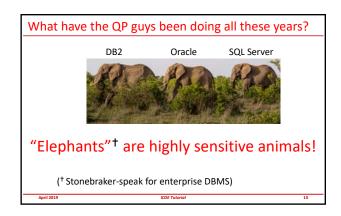


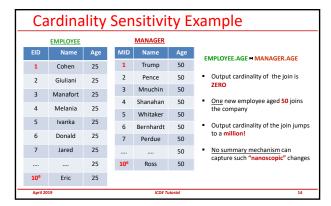






Run-time Sub-optimality The supposedly optimal plan-choice may actually turn out to be highly sub-optimal (e.g. a 1000 times worsel) when the query is executed with this plan. This adverse effect is due to errors in: (a) cost model Reasons: Simple linear models, operator-agnostic features, fixed coefficients, system dynamics ... (b) cardinality model Reasons: Coarse statistics, outdated statistics, attribute value independence (AVI) assumption, multiplicative error propagation, query construction, ...





Proof by Authority [Guy Lohman, IBM]



Snippet from April 2014 Sigmod blog post on "Is Query Optimization a "Solved" Problem?"

The root of all evil, the Achilles Heel of query optimization, is the estimation of the size of intermediate results, known as cardinalities. The cardinality model can easily introduce errors of many orders of magnitude! With such errors, the wonder isn't "Why did the optimizer pick a bad plan?" Rather, the wonder is "Why would the optimizer ever pick a decent plan?"

oril 2019

Sound-bites

- Little difference between worst-case and average-case in Query Processing
- It is far easier to win money at the Macau casinos than to get query processing right!

April 2019 ICDE Tutorial

Prior DB Research (lots!)

- · Sophisticated estimation techniques
 - SIGMOD 1999/2010, VLDB 2001/2009/2011, ..., CIDR 2019
 - e.g. wavelet histograms, self-tuning histograms, deep learning
- Selection of stable plans
 - SIGMOD 1994/2005/2010, PODS 1999/2002, VLDB 2008, ..., VLDB 2017
 - e.g. Variance-aware plan selection
- Runtime re-optimization techniques
 - SIGMOD 1998/2000/2004/2005, ..., arXiv 2019 [Stonebraker et al]
 - e.g. POP (progressive optimization), RIO (re-optimizer)

Several novel ideas and formulations, but are they robust?

April 2010

DE Tutorial

Is there any hope?

Over last decade, several promising advances that collectively promise to soon make robustness a contemporary reality – we will survey these techniques in the tutorial.

April 2019

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QP Robustness

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ROBUSTNESS DEFINITION

- Multiple perspectives, no consensus
 - If worst-case performance is improved at the expense of averagecase performance, is that acceptable?
 - Is it to be defined on a query instance basis, or "in expectation"?
- Ultimately, robustness definition is application dependent
- Graceful performance profile no "cliffs"
- Seamless scaling with workload complexity, database size, distributional skew, join correlations
- Provable guarantees on worst-case performance (relative to an offline ideal that makes all the right decisions)

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