

Optimizer Optimizer Implementation Part 3





Carnegie Mellon University

OBSERVATION

The best plan for a query can change as the database evolves over time.

- \rightarrow Physical design changes.
- \rightarrow Data modifications.
- \rightarrow Prepared statement parameters.
- \rightarrow Statistics updates.

The query optimizers that we have talked about so far all generate a plan for a query <u>before</u> the DBMS executes a query.

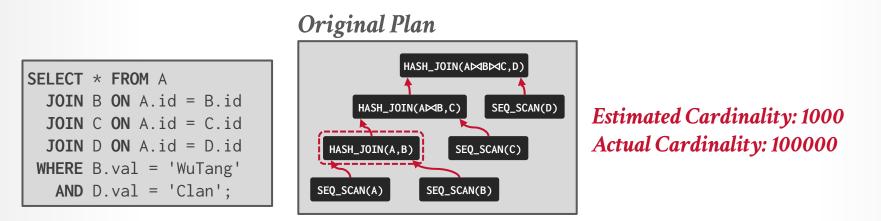
BAD QUERY PLANS

- The most common problem in a query plan is incorrect join orderings.
- \rightarrow This occurs because of inaccurate cardinality estimations that propagate up the plan.

If the DBMS can detect how bad a query plan is, then it can decide to <u>adapt</u> the plan rather than continuing with the current sub-optimal plan.



BAD QUERY PLANS



If the optimizer knew the true cardinality, would it have picked the same the join ordering, join algorithms, or access methods?



WHY GOOD PLANS GO BAD

Estimating the execution behavior of a plan to determine its quality relative to other plans.

These estimations are based on a <u>static</u> summarizations of the contents of the database and its operating environment:

- \rightarrow Statistical Models / Histograms / Sampling
- \rightarrow Hardware Performance
- \rightarrow Concurrent Operations



OPTIMIZATION TIMING

Choice #1: Static Optimization

- \rightarrow Select the best plan prior to execution.
- \rightarrow Plan quality is dependent on cost model accuracy.
- \rightarrow Can amortize over executions with prepared statements.

Choice #2: Dynamic Optimization

- \rightarrow Select operator plans on-the-fly as queries execute.
- \rightarrow Will have re-optimize for multiple executions.
- → Difficult to implement/debug (non-deterministic)

Choice #3: Adaptive Optimization

- \rightarrow Compile using a static algorithm.
- \rightarrow If the estimate errors > threshold, change or re-optimize.

ADAPTIVE QUERY OPTIMIZATION

Modify the execution behavior of a query by generating multiple plans for it:

- \rightarrow Individual complete plans.
- \rightarrow Embed multiple sub-plans at materialization points.

Use information collected during query execution to improve the quality of these plans.
→ Can use this data for planning one query or merge the it back into the DBMS's statistics catalog.



ADAPTIVE QUERY OPTIMIZATION

Approach #1: Modify Future Invocations

Approach #2: Replan Current Invocation

Approach #3: Plan Pivot Points

MODIFY FUTURE INVOCATIONS

The DBMS monitors the behavior of a query during execution and uses this information to improve subsequent invocations.

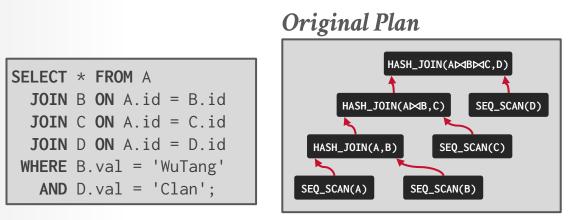
Approach #1: Plan Correction Approach #2: Feedback Loop

- The DBMS tracks the history of query invocations:
- \rightarrow Cost Estimations
- \rightarrow Query Plan
- \rightarrow Runtime Metrics

If the DBMS generates a new plan for a query, then check whether that plan performs worse than the previous plan.

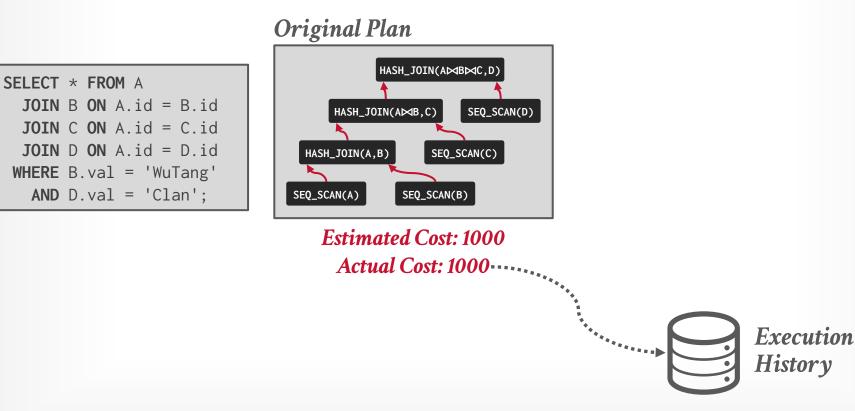
 \rightarrow If it regresses, then switch back to the cheaper plans.



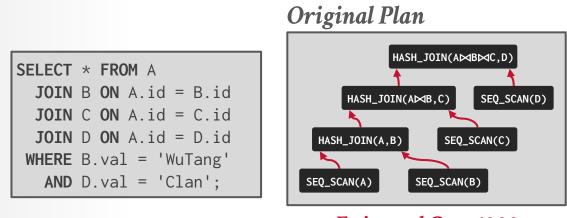


Estimated Cost: 1000 Actual Cost: 1000







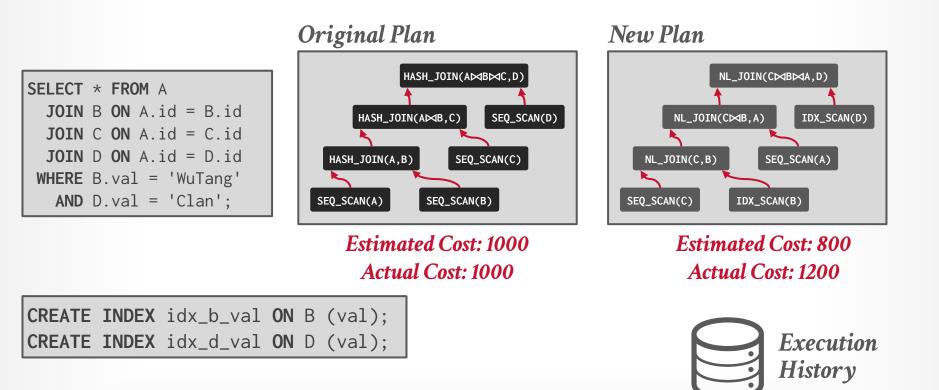


Estimated Cost: 1000 Actual Cost: 1000

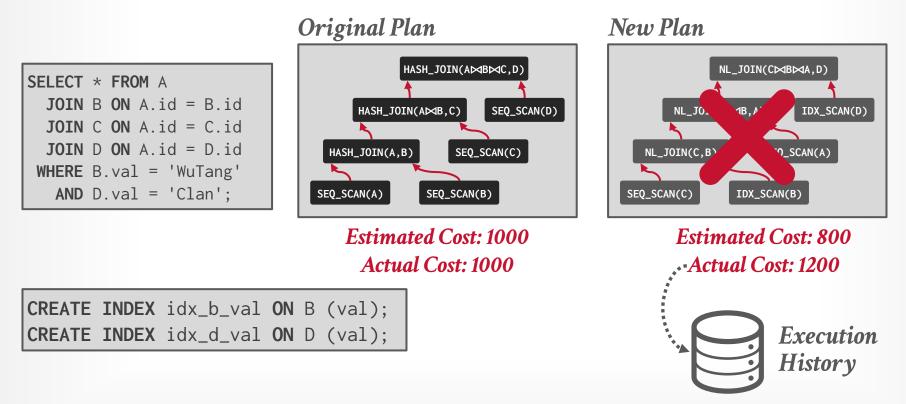
CREATE INDEX idx_b_val ON B (val); CREATE INDEX idx_d_val ON D (val);

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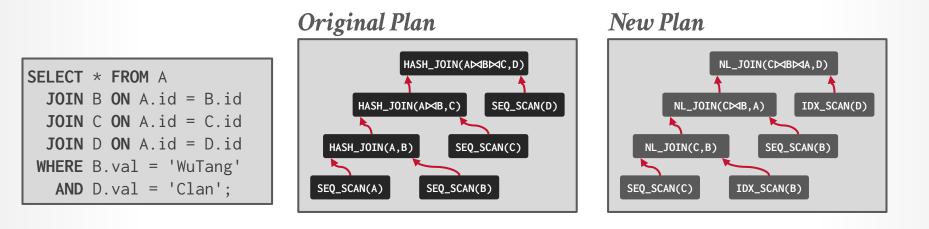
Combine useful sub-plans from queries to create potentially better plans.

- \rightarrow Sub-plans do not need to be from the same query.
- \rightarrow Can still use sub-plans even if overall plan becomes invalid after a physical design change.

Uses a dynamic programming search (bottom-up) that is not guaranteed to find a better plan.

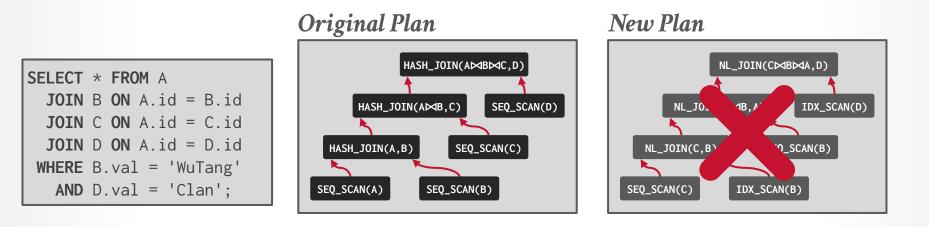






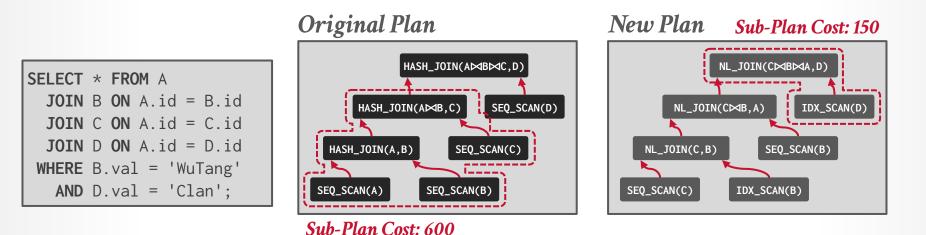
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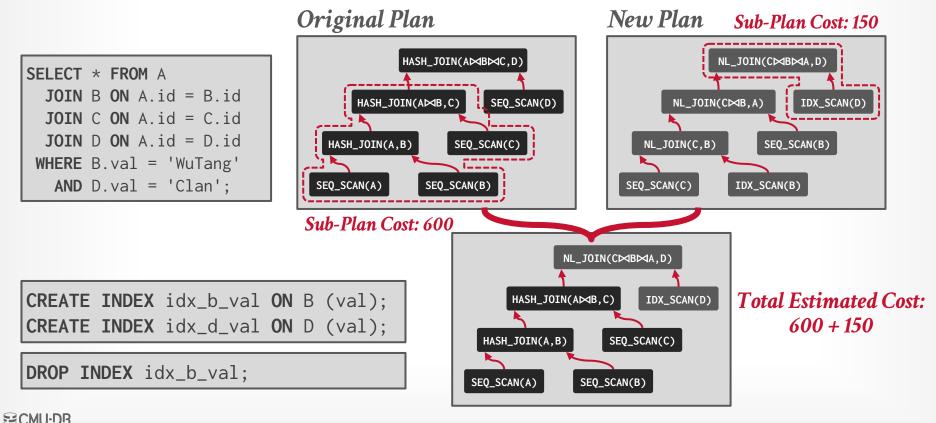
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DROP INDEX idx_b_val;



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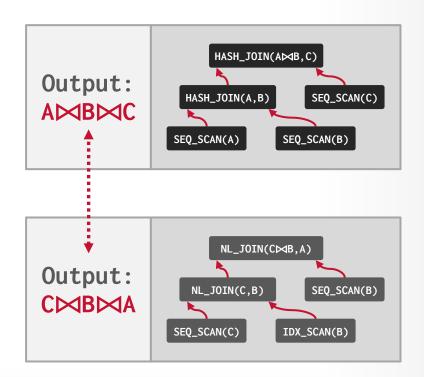
DROP INDEX idx_b_val;



IDENTIFYING EQUIVALENT SUBPLANS

Sub-plans are equivalent if they have the same logical expression and required physical properties.

Use simple heuristic that prunes any subplans that never be equivalent (e.g., access different tables) and then matches based on comparing expression trees.



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Generate a graph that contains all possible sub-plans.

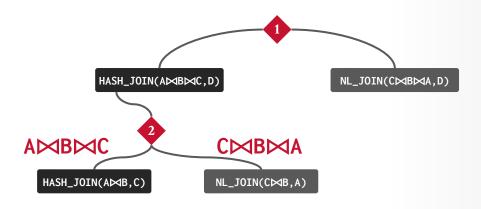


Add or operators to indicate alternative paths through the plan.

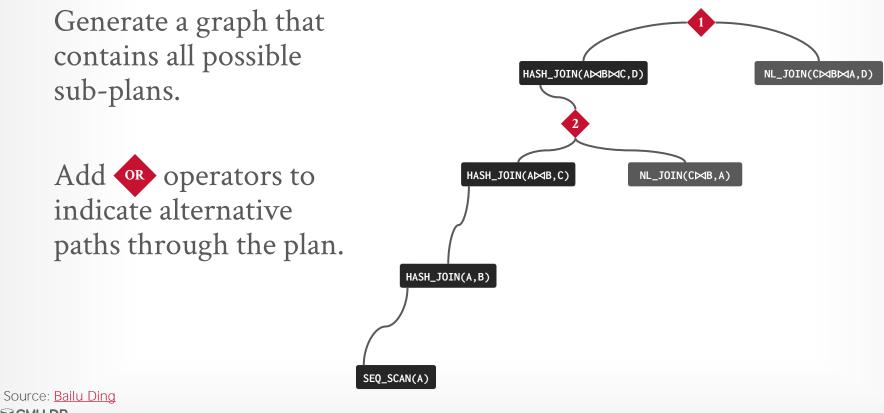
Source: Bailu Ding SCMU-DB 15-721 (Spring 2024)

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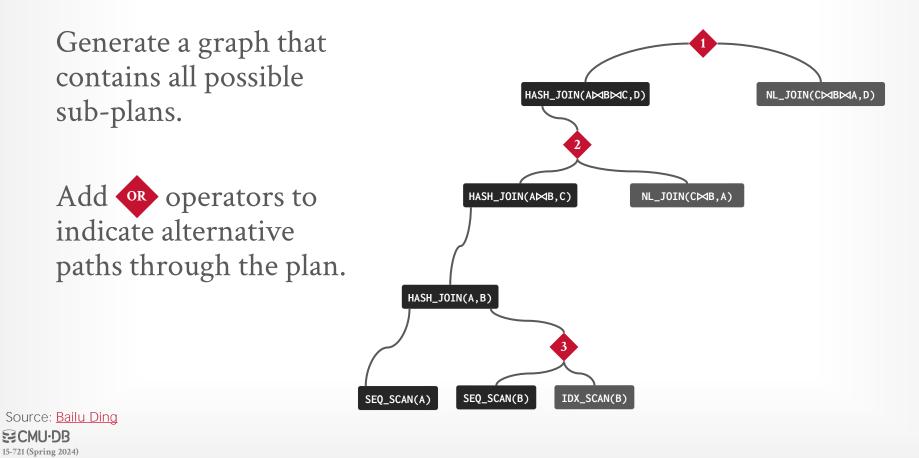
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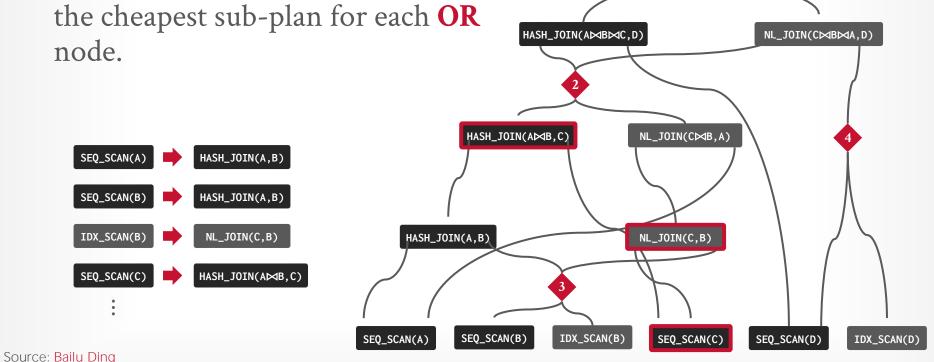


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CONSTRUCTING STITCHED PLANS

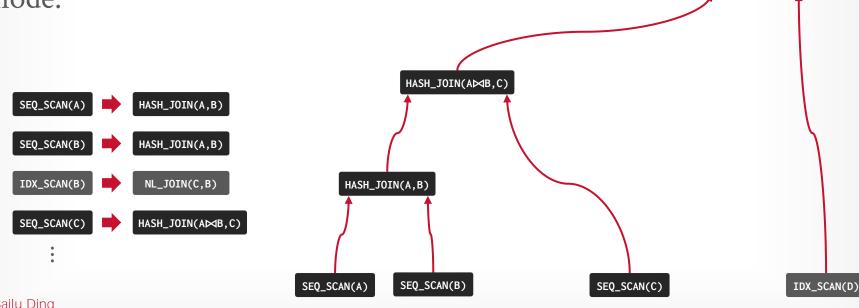




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CONSTRUCTING STITCHED PLANS

Perform bottom-up search that selects the cheapest sub-plan for each **OR** node.



Source: Bailu Ding

NL_JOIN(CMBMA,D)

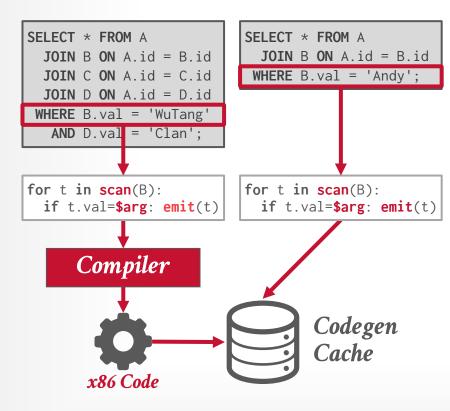
REDSHIFT: CODEGEN STITCHING

SELECT	*	FROM A
JOIN	В	ON A.id = B.id
JOIN	С	ON A.id = C.id
JOIN	D	ON A.id = D.id
WHERE	Β.	val = 'WuTang'
AND	D.	val = 'Clan';

Redshift is a transpilation-based codegen engine.

To avoid the compilation cost for every query, the DBMS caches subplans and then combines them at runtime for new queries.

REDSHIFT: CODEGEN STITCHING



SPCMU·DB 15-721 (Spring 2024) Redshift is a transpilation-based codegen engine.

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IBM DB2: LEARNING OPTIMIZER

Update table statistics as the DBMS scans a table during normal query processing.

Check whether the optimizer's estimates match what it encounters in the real data and incrementally updates them.



REPLAN CURRENT INVOCATION

If the DBMS determines that the observed execution behavior of a plan is far from its estimated behavior, them it can halt execution and generate a new plan for the query.

Approach #1: Start-Over from Scratch Approach #2: Keep Intermediate Results



```
CREATE TABLE fact (
  id INT PRIMARY KEY,
  dim1_id INT
   ♥ REFERENCES dim1 (id),
  dim2_id INT,
   ♥ REFERENCES dim2 (id)
);
        CREATE TABLE dim1 (
          id INT, val VARCHAR
        );
        CREATE TABLE dim2 (
```

First compute Bloom filters on dimension tables.

Probe these filters using fact table tuples to determine the ordering of the joins. 20

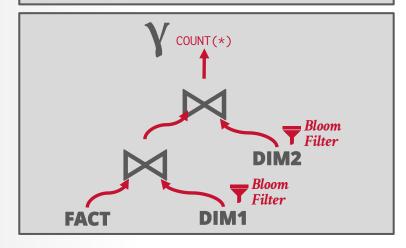
Only supports left-deep join trees on star schemas.



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id INT, val VARCHAR

SELECT COUNT(*) FROM fact AS f
JOIN dim1 ON f.dim1_id = dim1.id
JOIN dim2 ON f.dim2_id = dim2.id



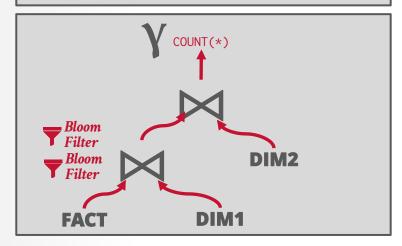
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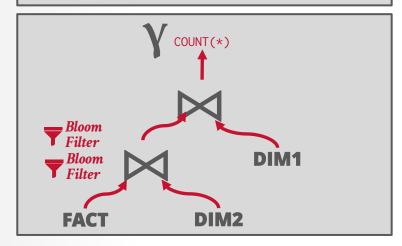
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OOKING AHEAD MAKES QUERY PLANS ROBUST

PLAN PIVOT POINTS

The optimizer embeds alternative sub-plans at materialization points in the query plan.

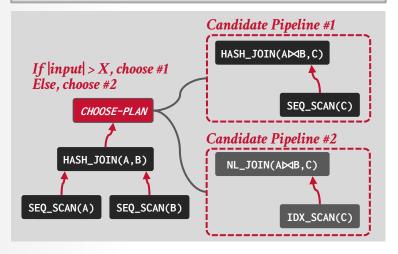
The plan includes "pivot" points that guides the DBMS towards a path in the plan based on the observed statistics.

Approach #1: Parametric Optimization Approach #2: Proactive Reoptimization



PARAMETRIC OPTIMIZATION

SELECT * FROM A
JOIN B ON A.id = B.id
JOIN C ON A.id = C.id;

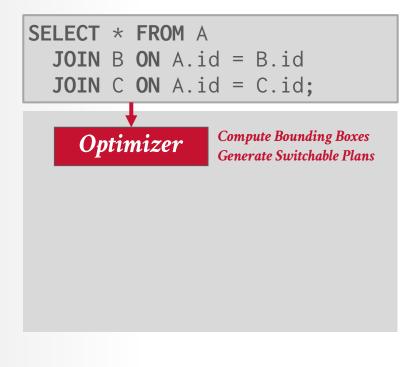


Generate multiple sub-plans per pipeline in the query.

Add a *choose-plan* operator that allows the DBMS to select which plan to execute at runtime.

First introduced as part of the Volcano project in the 1980s.



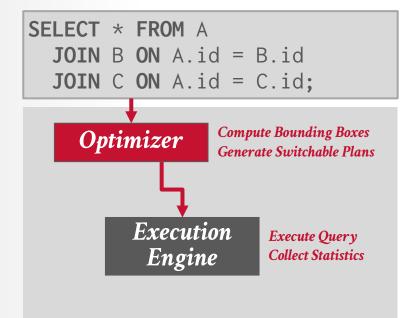


Generate multiple sub-plans within a single pipeline.

Use a *switch* operator to choose between different sub-plans during execution in the pipeline.

Computes bounding boxes to indicate the uncertainty of estimates used in plan.



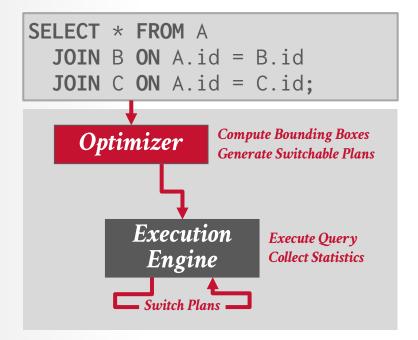


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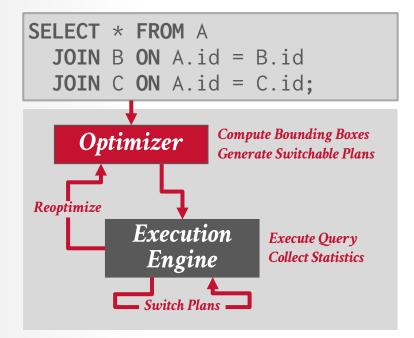




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PROACTIVE RE-OPTIMIZATION

SIGMOD 2005

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Generate multiple sub-plans within a single pipeline.

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Choice #1: Hints

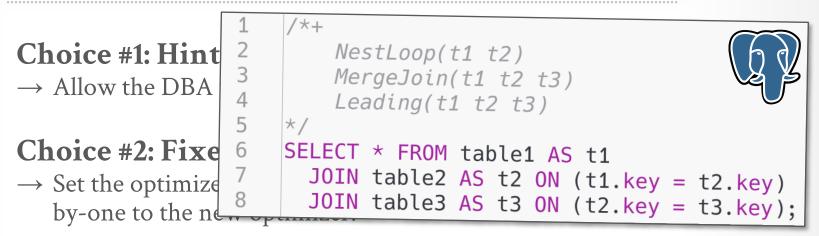
 \rightarrow Allow the DBA to provide hints to the optimizer.

Choice #2: Fixed Optimizer Versions

→ Set the optimizer version number and migrate queries oneby-one to the new optimizer.

Choice #3: Backwards-Compatible Plans

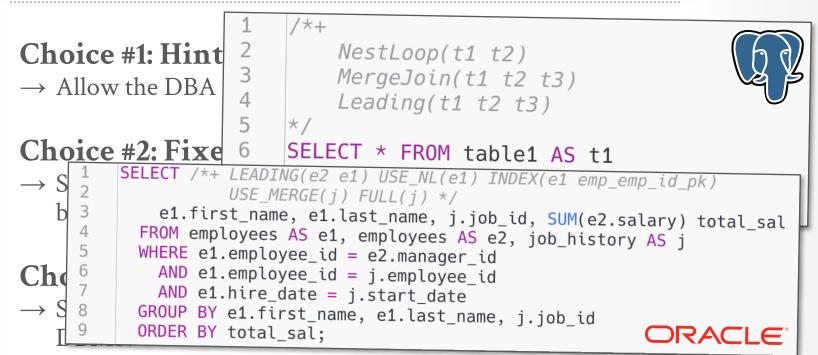
 \rightarrow Save query plan from old version and provide it to the new DBMS.



Choice #3: Backwards-Compatible Plans

 \rightarrow Save query plan from old version and provide it to the new DBMS.





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SECMU-DB 15-721 (Spring 2024) Choice #1: Hints SELECT * FROM dbo.tableA AS a Microsoft[®] INNER JOIN dbo.tableB AS b ON b.ID = a.ID Server 3 OPTION (\rightarrow Allow the DBA to t USE PLAN N'<?xml version="1.0" encoding="utf-16"?> <ShowPlanXML xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance" xmlns:xsd="http://</pre> www.w3.org/2001/XMLSchema" Version="1.518" Build="13.0.5026.0" xmlns="http:// schemas.microsoft.com/sqlserver/2004/07/showplan"> <BatchSequence> Choice #2: Fixed <Batch> 8 <Statements> <StmtSimple StatementCompId="1" StatementEstRows="1" StatementId="1"</pre> \rightarrow Set the optimizer v 4 StatementOptmLevel="FULL" StatementOptmEarlyAbortReason="GoodEnoughPlanFound" CardinalityEstimationModelVersion="70" StatementSubTreeCost="0.00657068" by-one to the new 4 StatementText="SELECT *

FROM dbo.tableA a

INNER LOOP JOIN dbo.tableB b

ON b.ID = a.ID" StatementType="SELECT" QueryHash="0x5126A10B217E55B6" QueryPlanHash="0x3700F7E4E3143DF3" RetrievedFromCache="false" SecurityPolicyApplied="false"> 4 10 <StatementSetOptions ANSI_NULLS="true" ANSI_PADDING="true"</pre> Choice #3: Backw ANSI WARNINGS="true" ARITHABORT="true" CONCAT_NULL_YIELDS_NULL="true" NUMERIC_ROUNDABORT="false" QUOTED_IDENTIFIER="true" /> <QueryPlan CachedPlanSize="16" CompileTime="0" CompileCPU="0" \rightarrow Save query plan frq₁₂ CompileMemory="152"> <MemoryGrantInfo SerialRequiredMemory="0" SerialDesiredMemory="0" /> <OptimizerHardwareDependentProperties DBMS. دے EstimatedAvailableMemoryGrant="1056000" EstimatedPagesCached="3168000" EstimatedAvailableDegreeOfParallelism="8" MaxCompileMemory="364530648" /> \hookrightarrow 14 <RelOp AvgRowSize="15" EstimateCPU="4.18E-06" EstimateIO="0" 4 EstimateRebinds="0" EstimateRewinds="0" EstimatedExecutionMode="Row" \hookrightarrow EstimateRows="1" LogicalOp="Inner Join" NodeId="0" Parallel="false" 4 PhysicalOp="Nested Loops" EstimatedTotalSubtreeCost="0.00657068"> 15 <OutputList> 16 <ColumnReference Database="[TestDB]" Schema="[dbo]" Table="[tableA]" Alias="[a]" Column="ID" />

PARTING THOUGHTS

The "plan-first execute-second" approach to query planning is notoriously error prone.

Optimizers should work with the execution engine to provide alternative plan strategies and receive feedback.

Adaptive techniques now appear in many of the major commercial DBMSs \rightarrow DB2, Oracle, MSSQL, TeraData



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NEXT CLASS

Let's understand how these cost models work and why they are so bad.

