Carnegie Mellon University

Database Query Optimization

Unnesting Queries

LAST CLASS

Parallelization of independent transformations in a top-down optimizer.

→ Another example of the need to track dependencies between parts of the query plan and optimization process.

This concludes the distinction between bottom-up and top-down methods.

SUBQUERIES

SQL allows a nested **SELECT** subquery to exist (almost?) anywhere in another query. \rightarrow Projection, **FROM**, **WHERE**, **LIMIT**, **HAVING**

 \rightarrow Results of the inner subquery are passed to the outer query.

Such nesting enables more expressive queries without having to use separate queries to prepare intermediate results.

Key Distinction: Uncorrelated vs. Correlated

An uncorrelated subquery does <u>not</u> reference any attributes from the (calling) outer query.

The DBMS only needs to <u>logically</u> execute the subquery once and reuse its result for all tuples in outer query. \rightarrow Most DBMSs will do this.

SELECT	name
FROM	students
WHERE	score =
(SELE	<pre>ECT MAX(score) FROM students);</pre>

A correlated subquery refers to one or more attributes from outside of the subquery (i.e., the outer query).

SELECT	name, major
FROM	students AS s1
WHERE	score =
	(SELECT MAX(s2.score)
	FROM students AS s2
	<pre>WHERE s2.major = s1.major);</pre>

name	major	score
GZA	CompSci	90
RZA	CompSci	80
ODB	Streets	100

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The DBMS logically evaluates the subquery on each tuple in the outer query because the result can change per tuple.

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name	major	score
GZA	CompSci	90
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s1.major='CompSci'

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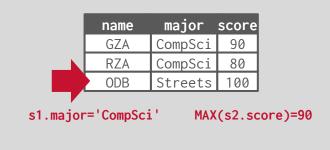
SELECT	name, major	
FROM	students AS s1	
WHERE	score =	
	(SELECT MAX(s2.score)	
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name	major	score
GZA	CompSci	90
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s1.major='CompSci'

A correlated subquery refers to one or more attributes from outside of the subquery (i.e., the outer query).

	name, major students AS s1	name GZA	major CompSci
	score =		
(SELECT MAX(s2.score)			
FROM students AS s2			
	WHERE s2.maj	or = s1.	<pre>major);</pre>



A correlated subquery refers to one or more attributes from outside of the subquery (i.e., the outer query).

The DBMS logically evaluates the subquery on each tuple in the outer query because the result can change per tuple.

SELECT	name, major	name	major
FROM	students AS s1	GZA	CompSci
WHERE	score =		
(SELECT MAX(s2.score)			
FROM students AS s2			
	WHERE s2.maj	or = s1.	<pre>major);</pre>

name	major	score
GZA	CompSci	90
RZA	CompSci	80
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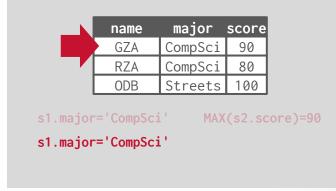
s1.major='CompS

AX(s2.score)=90

s1.major='CompSci'

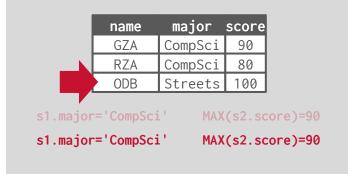
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	name, major students AS s1	name major GZA CompSci	
	score =		
(SELECT MAX(s2.score)			
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	name, major students AS s1	name GZA	major CompSci
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SELECT	name, major	name	major
FROM	students AS s1	GZA	CompSci
WHERE	score =		
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	WHERE s2.maj	or = s1.	<pre>major);</pre>

	name	major	score	
	GZA	CompSci	90	
	RZA	CompSci	80	
	ODB	Streets	100	
s1.major='CompSci' MAX(s2.score)=90 s1.major='CompSci' MAX(s2.score)=90				
s1.major='Streets'				

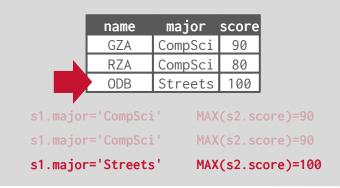
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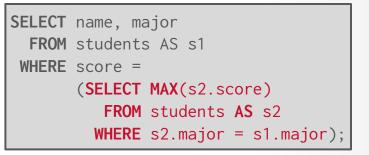
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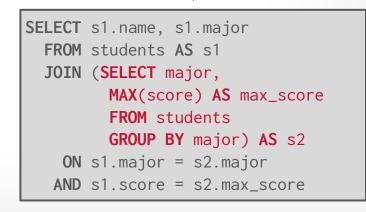
SELECT	name, major	name	major	
FROM	students AS s1	GZA	CompSci	
	score =	ODB	Streets	
(SELECT MAX(s2.score)				
FROM students AS s2				
WHERE s2.major = s1.major);				



The goal is for the optimizer to pull a correlated subquery up from an inner nesting level so that the DBMS can execute it as a join.

The optimizer needs to handle any amount of subquery nesting in any part of the query where it is allowed.





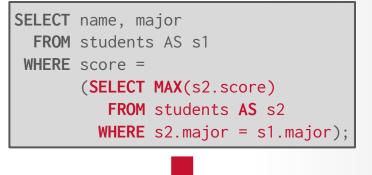
TODAY'S AGENDA

Binding Heuristic Rewriting German-style Unnesting (2015) German-style Unnesting (2025)

SUBQUERY BINDING

If you think of a subquery like a function call, then any column that can be passed to a function should be available to the subquery.

This can be challenging if the referenced columns are ambiguous.



SELECT	name, major
FROM	students AS s1
WHERE	<pre>score = subquery(s1.major);</pre>

SUBQUERY BINDING

SELECT:

- \rightarrow Normal columns
- \rightarrow AGGREGATE/GROUP columns

WHERE / GROUP BY:

 \rightarrow Any normal column available

HAVING:

→ AGGREGATE/GROUP columns

ORDER BY:

 \rightarrow Anything that can go in the root of **SELECT**.

LIMIT:

 \rightarrow No correlated columns allowed.

Source: Mark Raasveldt

```
SELECT (SELECT SUM(i1.i))
FROM integers AS i1;
```

SELECT subquery(SUM(i1.i))
FROM integers AS i1;

SELECT subquery(i1.i)
FROM integers AS i1;

SUBQUERY BINDING

SELECT:

- \rightarrow Normal columns
- → AGGREGATE/GROUP columns

WHERE / GROUP BY:

 \rightarrow Any normal column available

HAVING:

→ AGGREGATE/GROUP columns

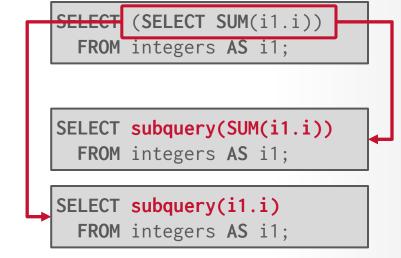
ORDER BY:

 \rightarrow Anything that can go in the root of **SELECT**.

LIMIT:

 \rightarrow No correlated columns allowed.

Source: Mark Raasveldt



HEURISTIC REWRITING

Since the early 1980s, optimizers relied on heuristics to identify specific query plan patterns to decorrelate nested subqueries.

The optimizer developer human codifies the patterns to look for when and how to decorrelate subqueries.

SOLite

11. Subguery Flattening

When a subguery occurs in the FROM clause of a SELECT, the simplest behavior is to evaluate the subguery into a transient table, then run the outer SELECT against the transient table. Such a plan can be suboptimal since the transient table will not have any indexes and the outer query (which is likely a join) will be forced to do a full table scan on the transient table.

To overcome this problem, SQLite attempts to flatten subqueries in the FROM clause of a SELECT. This involves inserting the FROM clause of the subquery into the FROM clause of the outer query and rewriting expressions in the outer query that refer to the result set of the subquery. For example:

SELECT t1.s, t2.b FROM t2, (SELECT x+y AS a FROM t1 WHERE z<100) WHERE a>5

Would be rewritten using query flattening as:

SELECT t1.x+t1.y AS m. t2.b FROM t2. t1 WHERE z<100 AND m>5

There is a long list of conditions that must all be met in order for query flattening to occur. Some of the constraints are marked as obsolete by italic text. These extra constraints are retained in the documentation to preserve the numbering of the other constraints.

Casual readers are not expected to understand all of these rules. A key take-away from this section is that the rules for determining if query flatting is safe or unsafe are subtle and complex. There have been multiple bugs over the years caused by over-aggressive query flattening. On the other hand, performance of complex gueries and/or gueries involving views tends to suffer if guery flattening is more conservative.

1. (Obsolete. Query flattening is no longer attempted for aggregate subqueries.)

2. (Obsolete. Query flattening is no longer attempted for aggregate subqueries.)

3. If the subquery is the right operand of a LEFT JOIN then

a, the subguery may not be a join, and b. the FROM clause of the subquery may not contain a virtual table, and c. the outer query may not be an aggregate.

4. The subguery is not DISTINCT.

- 5 (Subsumed into constraint 4)
- 6. (Obsolete. Query flattening is no longer attempted for aggregate subqueries.) 7. The subquery has a FROM clause.

- 8. The subquery does not use LIMIT or the outer query is not a join. 9. The subquery does not use LIMIT or the outer query does not use aggregates
- 10. (Restriction relaxed in 2005)
- 11. The subguery and the outer guery do not both have ORDER BY clauses
- 12 (Subsumed into constraint 3)
- 13. The subquery and outer query do not both use LIMIT.
- 14. The subquery does not use OFFSET.
- 15. If the outer query is part of a compound select, then the subquery may not have a LIMIT clause.
- 16. If the outer guery is an aggregate, then the subguery may not contain ORDER BY.
- 17. If the sub-query is a compound SELECT, then

a. all compound operators must be UNION ALL, and b. no terms with the subquery compound may be aggregate or DISTINCT, and c. every term within the subquery must have a FROM clause, and

d. the outer query may not be an aggregate, DISTINCT query, or join.

The parent and sub-query may contain WHERE clauses. Subject to rules (11), (12) and (13), they may also contain ORDER BY, LIMIT and OFFSET clauses.

- 18. If the sub-query is a compound select, then all terms of the ORDER by clause of the parent must be simple references to columns of the sub-query.
- 19. If the subquery uses LIMIT then the outer query may not have a WHERE clause. 20. If the sub-guery is a compound select, then it must not use an ORDER BY clause.
- 21. If the subguery uses LIMIT, then the outer query may not be DISTINCT.
- 22. The subguery may not be a recursive CTE.
- 23. (Subsumed into constraint 17d.)
- 24. (Obsolete. Query flattening is no longer attempted for aggregate subqueries.)

Query flattening is an important optimization when views are used as each use of a view is translated into a subguery.

Small Fast Reliable

Choose any three

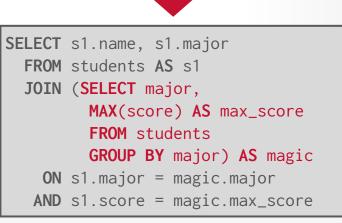
MAGIC SETS

Early technique for rewriting queries to include auxiliary "magic" tables that act as filters to reduce the amount of data processed during query execution.

Move correlated subqueries out of **WHERE** clause and into **FROM** clause.



SELECT name, major
FROM students AS s1
WHERE score =
 (SELECT MAX(s2.score)
 FROM students AS s2
 WHERE s2.major = s1.major);



MSSQL HEURISTICS

Use a set of small, independent, and orthogonal optimizations that collectively remove correlated subqueries.

Remove correlations by rewriting APPLY operators into standard relational algebra operators like outer joins.

$R \ \mathcal{A}^{\otimes} \ E$	Π	$R \otimes_{ ext{true}} E,$	(1)
if no par	ame	eters in E resolved from R	
$R \mathcal{A}^{\otimes} (\sigma_p E)$	=	$R \otimes_p E,$	(2)
if no parameters in E resolved from R			
$R \mathcal{A}^{\times} (\sigma_p E)$	=	$\sigma_p(R \mathrel{\mathcal{A}^{\times}} E)$	(3)
$R \ \mathcal{A}^{ imes} \ (\pi_v E)$	=	$\pi_{v \cup \operatorname{columns}(R)}(R \mathcal{A}^{\times} E)$	(4)
$R \mathcal{A}^{ imes} (E_1 \cup E_2)$	=	$(R \mathcal{A}^{\times} E_1) \cup (R \mathcal{A}^{\times} E_2)$	(5)
$R \mathcal{A}^{\times} (E_1 - E_2)$	=	$(R \mathcal{A}^{\times} E_1) - (R \mathcal{A}^{\times} E_2)$	(6)
$R \mathcal{A}^{\times} (E_1 \times E_2)$	=	$(R \mathcal{A}^{\times} E_1) \bowtie_{R.key} (R \mathcal{A}^{\times} E_2)$)(7)
$R \mathcal{A}^{\times} (\mathcal{G}_{A,F}E)$	=	$\mathcal{G}_{A \cup \operatorname{columns}(R),F}(R \ \mathcal{A}^{\times} \ E)$	(8)
$R \mathcal{A}^{\times} (\mathcal{G}_F^1 E)$	=	$\mathcal{G}_{\operatorname{columns}(R),F'}(R \mathcal{A}^{\operatorname{LOJ}} E)$	(9)



HEURISTIC REWRITING

Advantages:

- \rightarrow Transformed queries are more efficient.
- \rightarrow Decision to decorrelate can be a cost-based decision.
- \rightarrow Easy to control decorrelation by enabling/disabling rules.

Disadvantages:

- \rightarrow Hard to write rules for all possible correlations scenarios.
- \rightarrow Changing a small part of a query can make rules ineffective
- \rightarrow Maintaining transformation rules is a difficult.
- \rightarrow Handling all edge cases is exceedingly difficult.

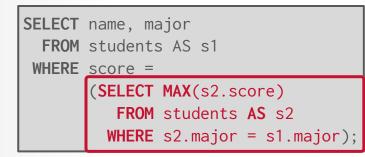
GERMAN-STYLE UNNESTING (2015)

Bottom-up method to eliminate dependent joins one-at-a-time by manipulating the query plan at the algebra level until the join's RHS no longer depends on the LHS.

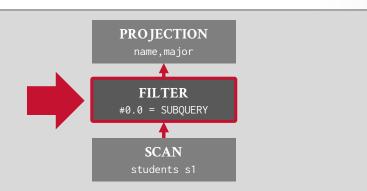
The optimizer then converts dependent joins to regular joins.

 \rightarrow Some queries switch from a $O(n^2)$ nested-loop join to a O(n) hash join.

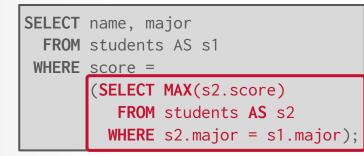




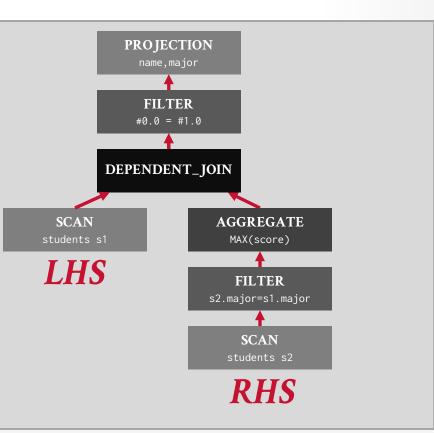
Introduce a <u>dependent join</u> logical operator to execute RHS once for every tuple in LHS.

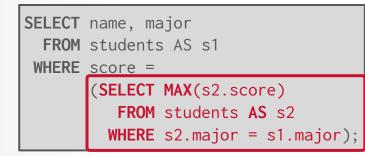


Source: Mark Raasveldt

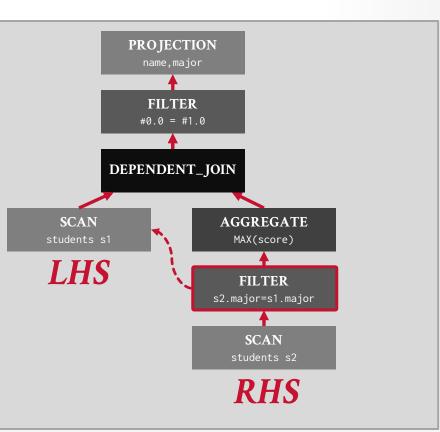


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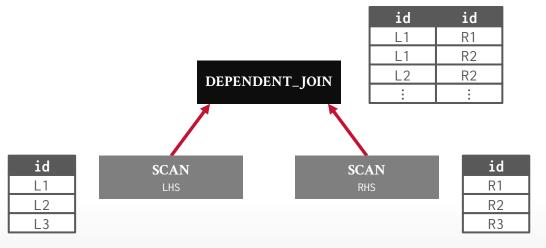
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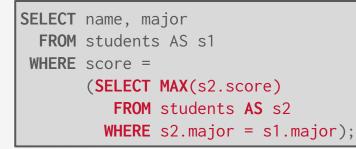
DEPENDENT JOIN

New **<u>dependent join</u>** relational algebra operator that denotes a correlated subquery.

- \rightarrow Evaluate RHS of the join for every tuple on the LHS.
- \rightarrow The operator combine results from every execution and return them as its output.



Source: Mayank Baranwal

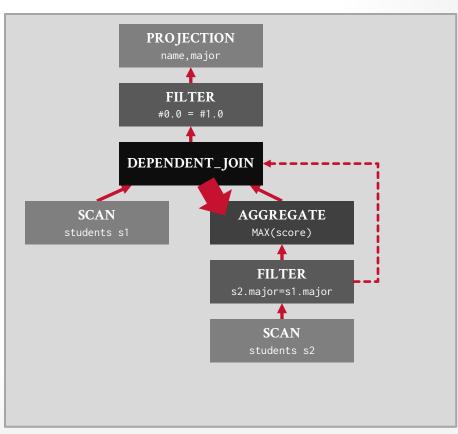


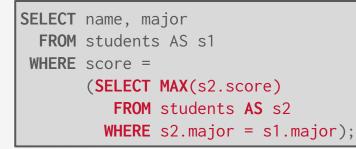
Push dependent join down into the RHS of the plan.

Only need to execute RHS once for every unique combination of correlated columns.

 \rightarrow Duplicate Elimination Scan

Source: Mark Raasveldt



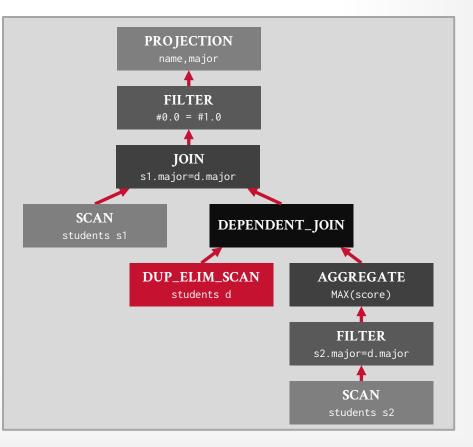


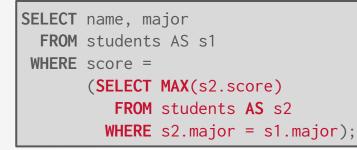
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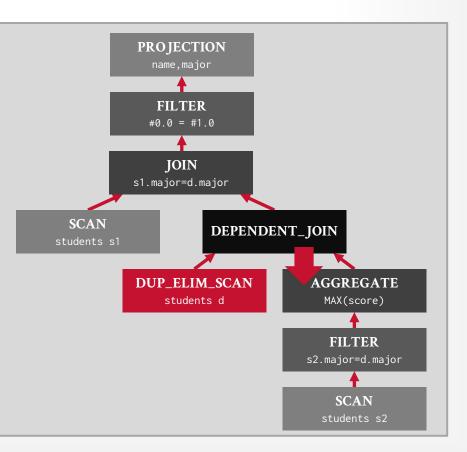
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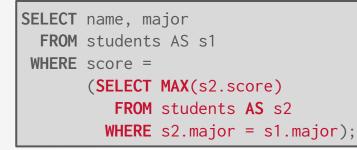
Source: Mark Raasveldt



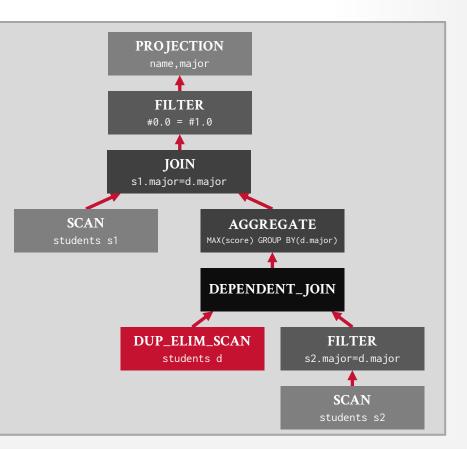


Keeping pushing dependent join as far down into the plan as is possible.

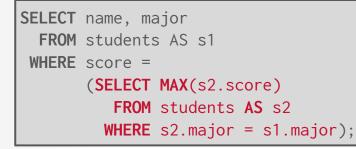




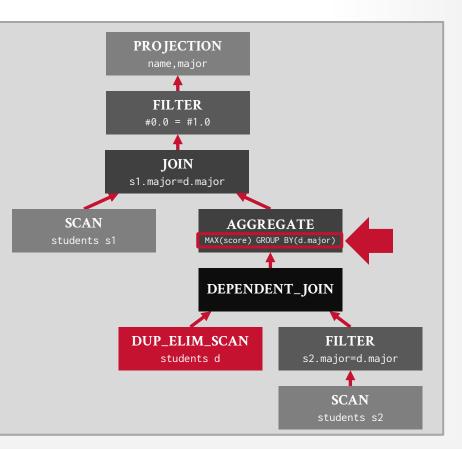
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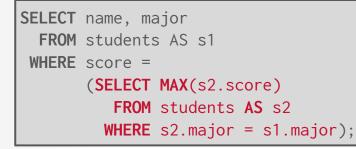
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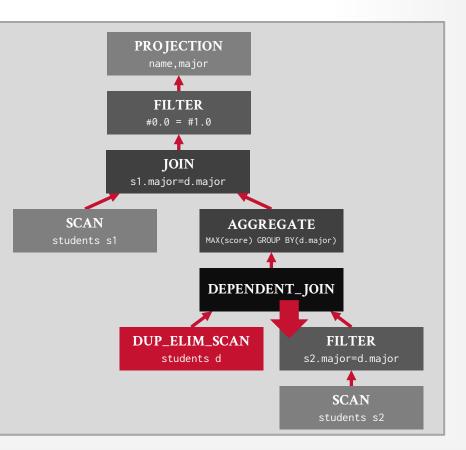
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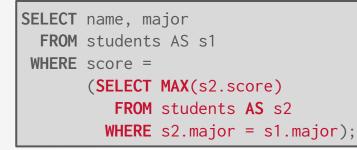
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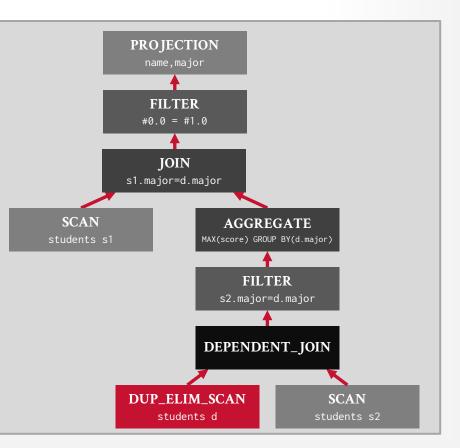
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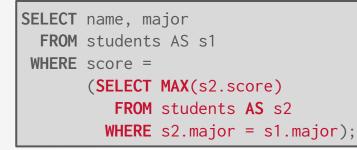


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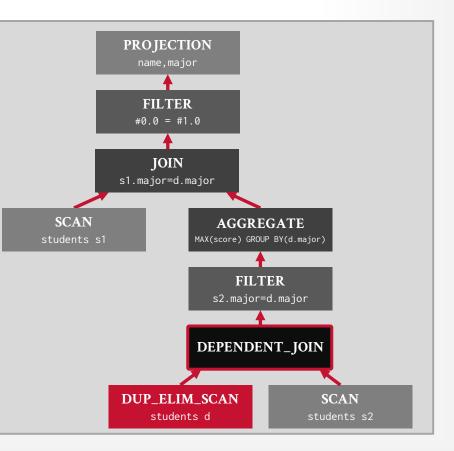


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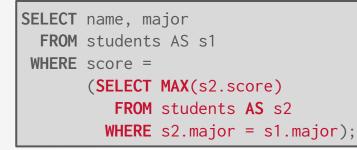




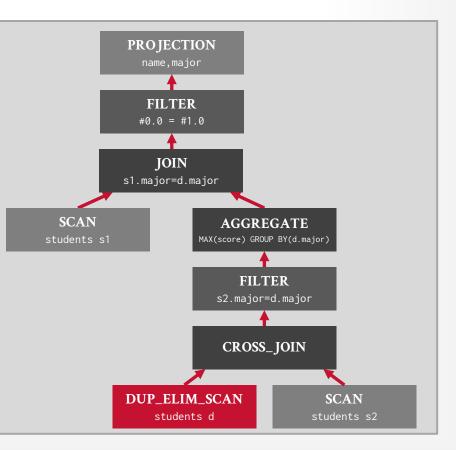
Convert the <u>dependent join</u> operator into a <u>cross join</u>.

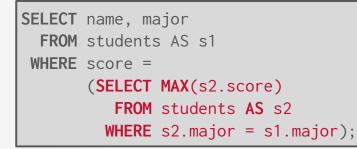


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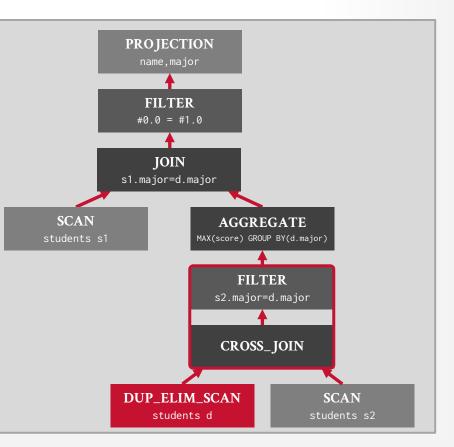
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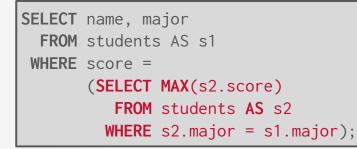


Convert the <u>dependent join</u> operator into a <u>cross join</u>.

Then convert the <u>cross join</u> into an <u>inner join</u>.

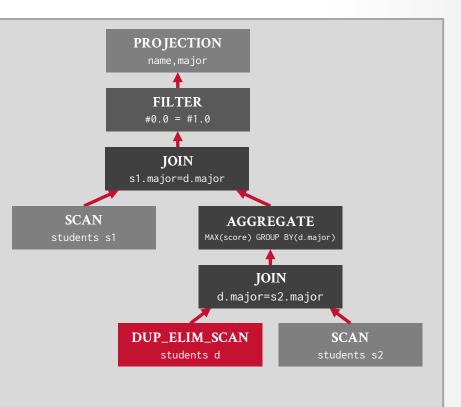


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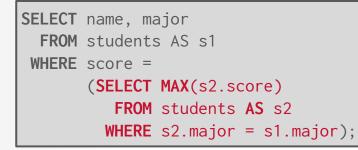


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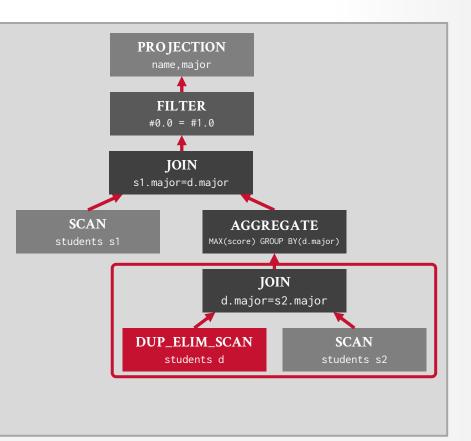


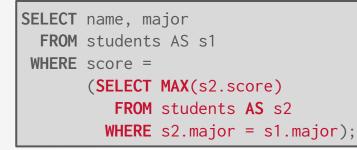
Source: Mark Raasveldt



Remove duplicate elimination scan entirely.

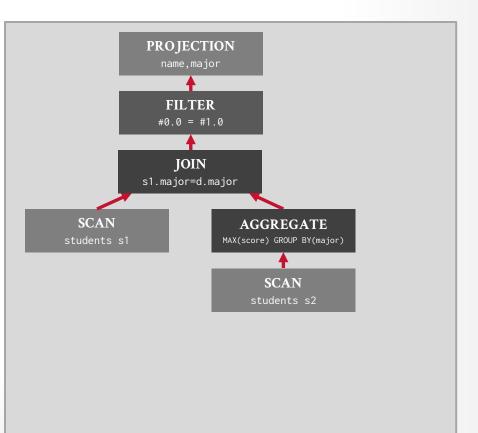
Remove the filter above the new join.

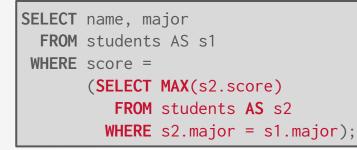




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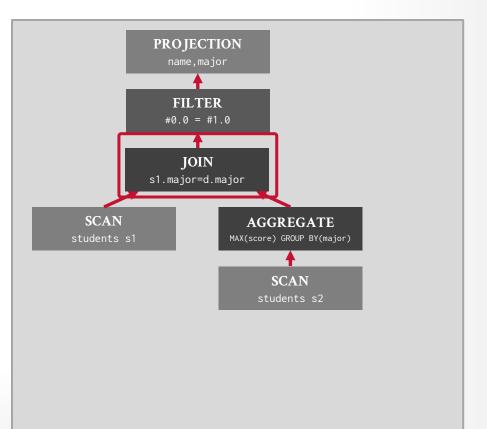
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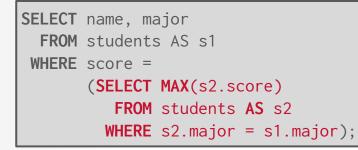




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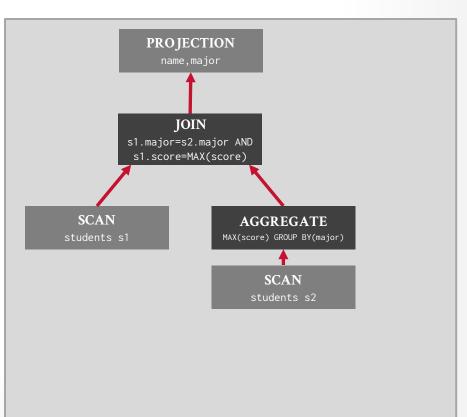
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```
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```



22

OBSERVATION

The 2015 unnesting approach handles most queries.
 → Known implementations in HyPer, Umbra, DuckDB, and DataBricks (partial).

But for queries with multiple nested dependent subqueries where rewriting to remove each dependent join one at a time leads to inefficient query plans.

OBSERVATION

The 2015 2.3 Limitations of the Bottom-Up Approach \rightarrow Known While the bottom-up approach handles most queries just fine, it unfortunately degenerates DataBr in some corner cases. We were originally notified about this by Sam Arch, who translated complex UDFs into pure SQL [Fr24]. There, similar to our original example in Figure 1, it could happen that dependent subqueries are nested inside each other. We show a variation² But for d subqueries where rewriting to remo-Similarly, for the original query from Sam Arch that motivated this work, which is procbench UDF Query 18 [GR21] after passing through Apfel [FHG22]: The unnesting strategy from [NK15] leads to memory exhaustion, while with our new top-down unnesting Umbra answers the query in 251ms on TPC-DS SF1.

OBSERVATION

The 2015 2.3 Limitations of the Bottom-Up Approach \rightarrow Known

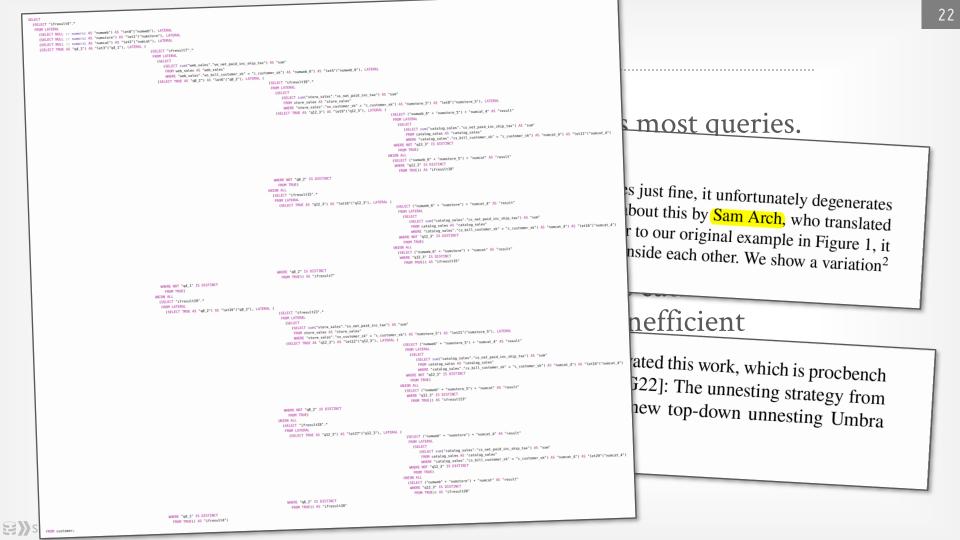
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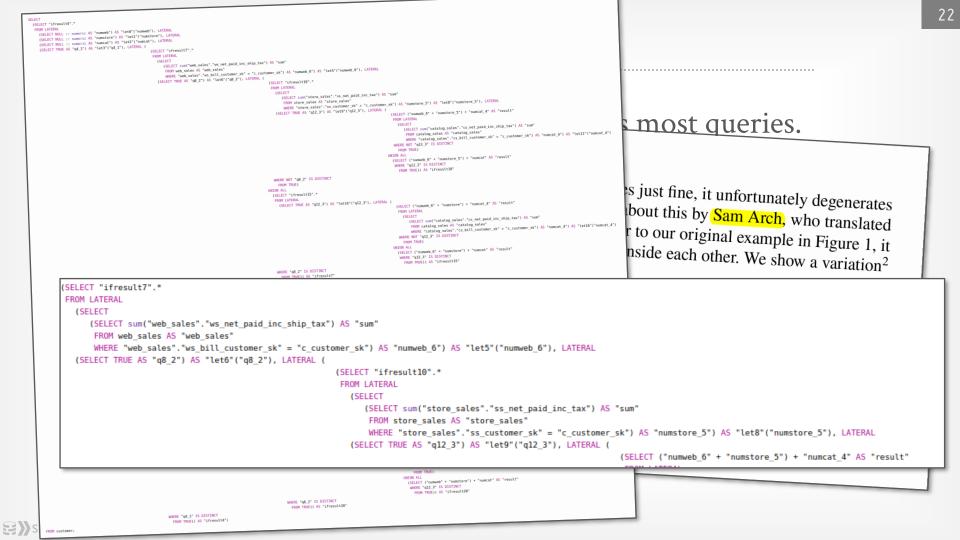
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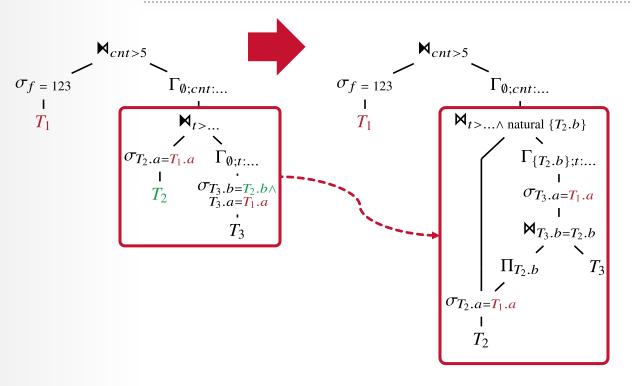
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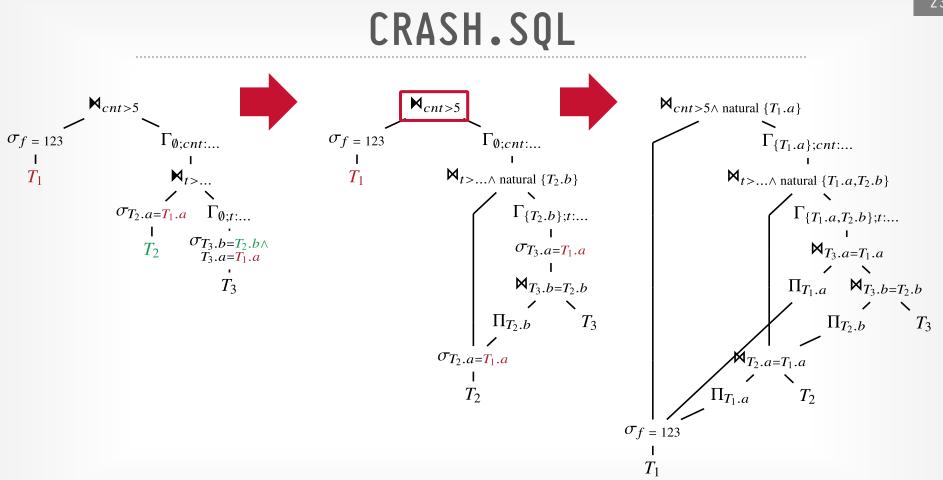




CRASH.SQL

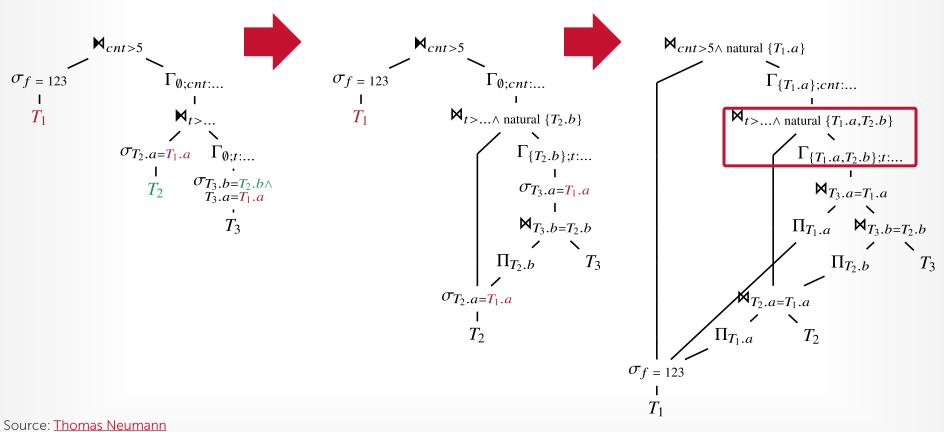


Source: Thomas Neumann

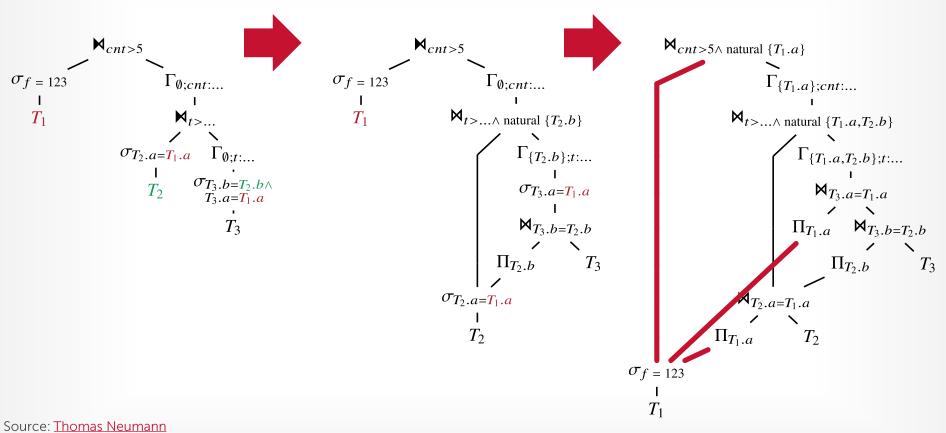


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HOLISTIC UNNESTING (2025)

Remove all dependent joins at the same time starting at the top of the query plan.

- → Keep track of where they are in the plan and then rewrite all operators in a top-down pass until each join is unnecessary or it can be safely added.
- \rightarrow Avoids pushing dependency sets across joins.

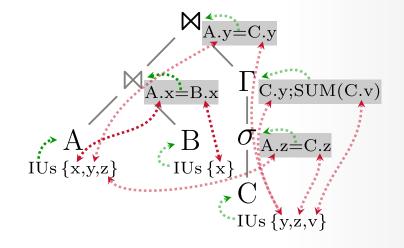
The optimizer needs an efficient way to identify the flow of attributes through the plan...

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INDEXED ALGEBRA

Unnesting subqueries requires the optimizer to reason about the dependencies and flow of attributes in a query plan's operators.

Maintain an auxiliary index of operator meta-data to facilitate faster examination of plans and to identify rewrite opportunities.



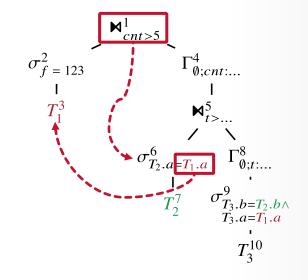
Operator Connections
Source Operator
IU Consumers



HOLISTIC UNNESTING: IDENTIFICATION

Identify dependent joins where the RHS accesses attributes provided by the LHS.

For each column accessed, compute the <u>lowest common ancestor</u> of operator o_1 that accesses a column and operator o_2 that provides the column. \rightarrow If $o_1 \neq o_2$, then it is a dependent join.

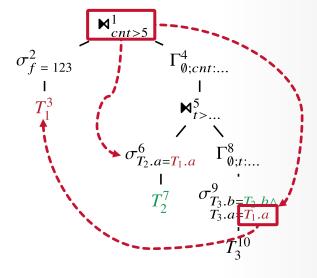


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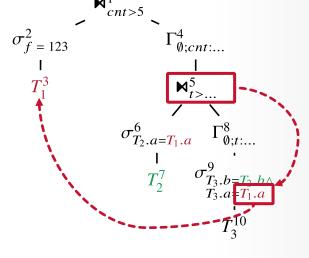
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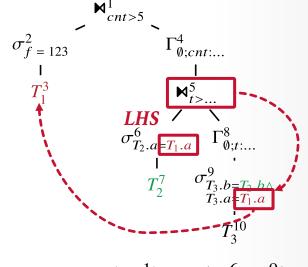
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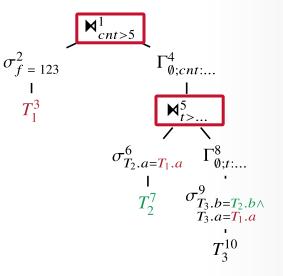
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SIMPLE ELIMINATION

Inspect all operators that access the LHS of a dependent join.

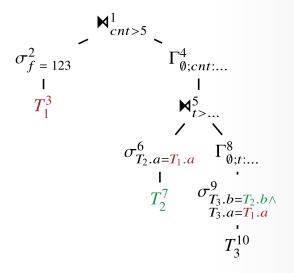
Then use the "simple" dependent join elimination discussed earlier. \rightarrow Move operators up towards the join.

Otherwise, use the full unnesting algorithm...

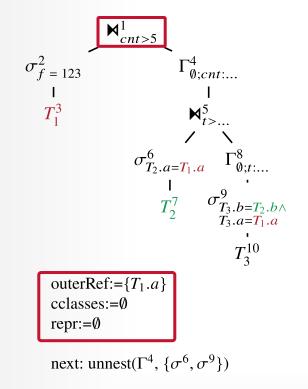


- Rewrite RHS of dependent join such that no references from the "outer" side occur anymore.
- \rightarrow Columns from the LHS that are accessed from the RHS.

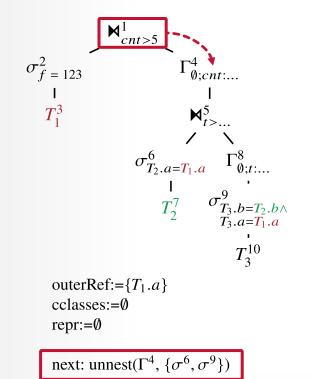
Maintain state about the algorithm's progress to keep track of where columns are coming from in plan.



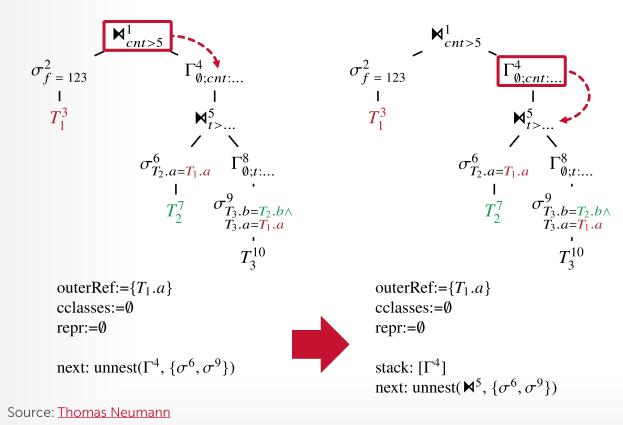
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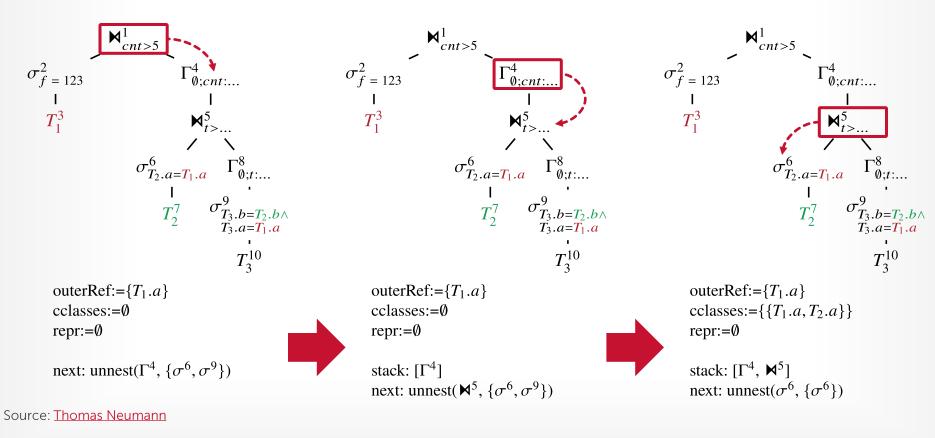


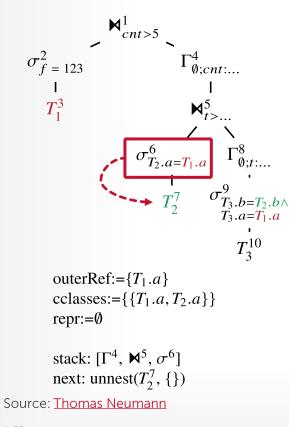
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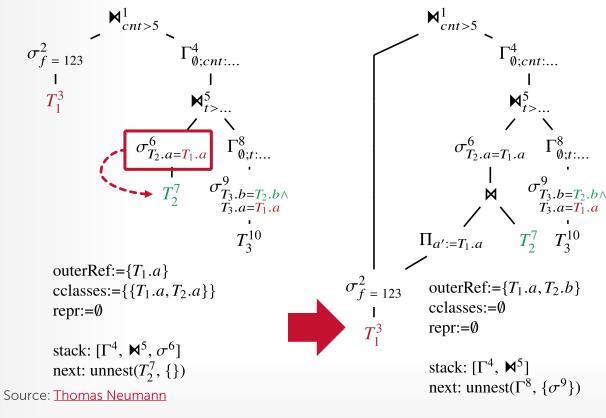


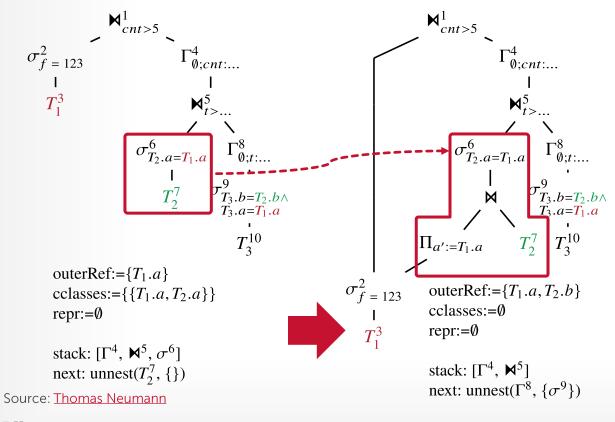
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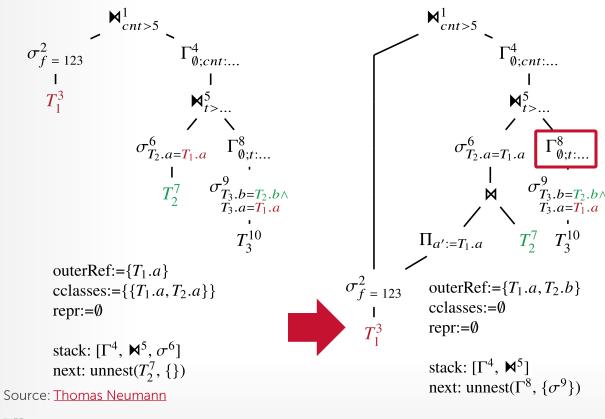


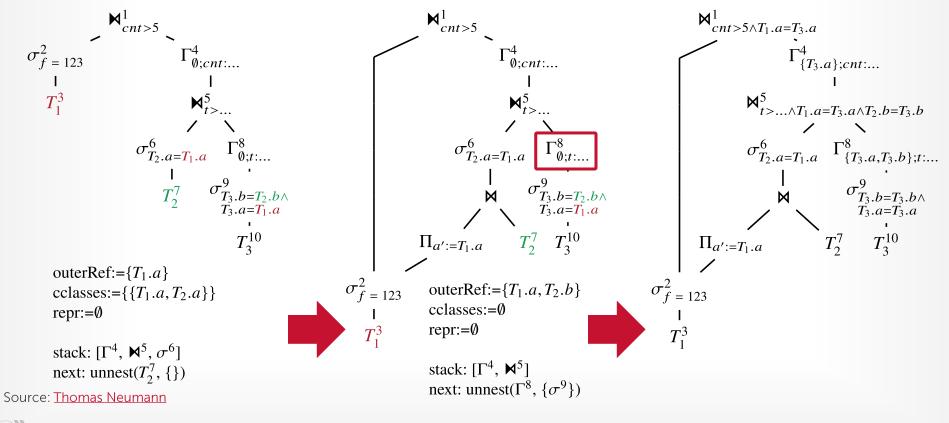


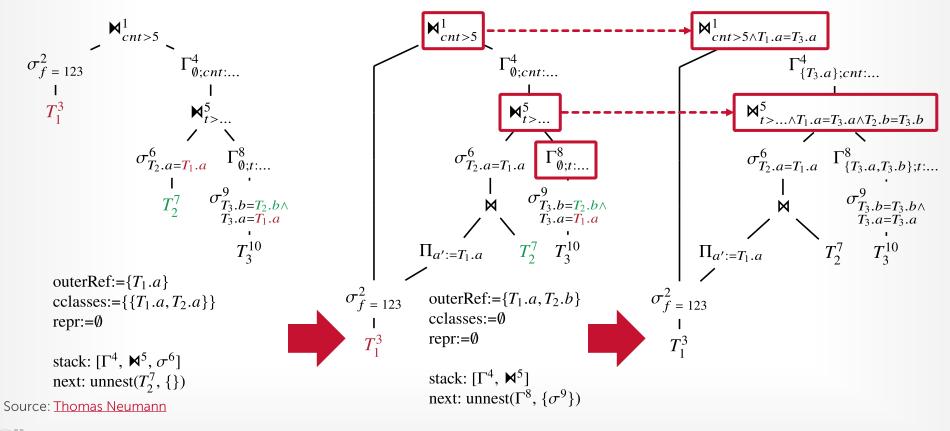












PARTING THOUGHTS

Holistic unnesting is the definitive way to decorrelate subqueries.

- \rightarrow Relies on DBMS supporting DAG query plans.
- \rightarrow Build indexes to speed up query plan analysis during optimization phases.

We will see correlated subqueries again when discussing UDF inlining.

NEXT CLASS

Cost Models! Statistics!

 \rightarrow aka when everything falls apart...