



DrillBeyond: Processing Multi-Result Open World SQL Queries

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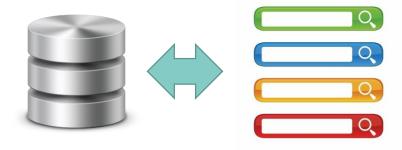
Motivation



DOMAINS OF...

- …information retrieval and
- database systems...

...HAVE BEEN TRADITIONALLY KEPT SEPARATE



DIFFERENCES IN...

•	type of data that is managed	(structured	versus	unstructured)
•	query language used	(fully specified query	versus	keyword query)
•	nature of the query result	(exact single answer	versus	ranked list of possible answers)
	usage scenarios	(analytic scenarios	versus	information gathering)

WE EXPLORE A NEW WAY...

...of merging the two paradigms for use in ad-hoc and self-service analytics



DrillBeyond



DRILLBEYOND = NOVEL HYBRID DBMS/IR SYSTEM

Blurring the lines between type of data managed, query language used, and nature of the query result

```
select n_name, gdp , avg(o_totalprice)
from nation, customer, orders
where
    n_nationkey=c_nationkey
    andc_custkey=o_custkey
    and gdp > 10.0
group by n_name, gdp
order by gdp desc
```

MIX OF...

- ...relational queries on a database and...
- ...top-k keyword-based searches (= Enity Augmentation Queries)

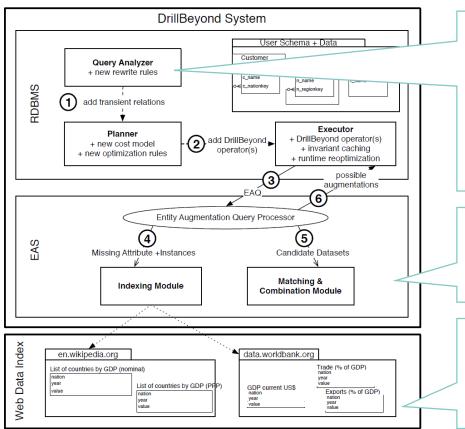






DrillBeyond – System Architecture





QUERY ANALYZER

- Maps SQL query tokens to the database catalog
- For unrecognized tokens (e.g. "gdp") we introduce transient metadata
- Query is rewritten to include an additional join with a transient relation

ENTITY AUGMENTATION SYSTEM

Implements the entity augmentation processing

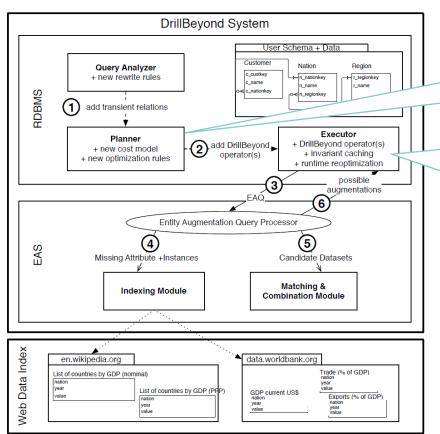
WEB DATA INDEX

- Index of Web tables
- Generic system exposing an interface for keyword-based document search

talk yesterday

DrillBeyond – System Architecture (2)





QUERY PLANNER

 Minimize the overhead of creating multiple result variants

EXECUTOR

• Implements the repeated execution of operator trees using the DrillBeyond operator ω



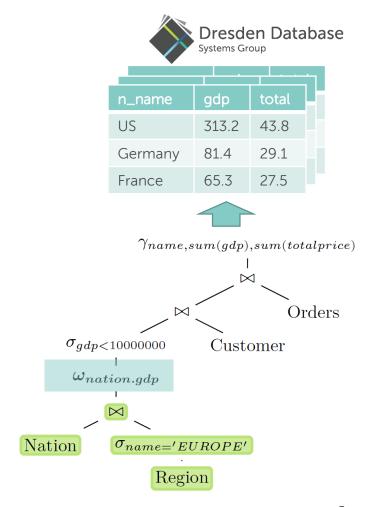
System Integration Challenges

MULTI-SOLUTION QUERY PROCESSING

- Top-k query result versions for a single user query
- Naïve way: process the query k times → query runtime increased by factor k
- Goals:
 - Minimizes duplicate work between query executions
 - Maximize invariant parts of multiple executions

QUERY PLANNING

- Web tables are not fully known at plan-time
 - → no selectivity information or data types at plan-time
- Goal: plan queries with relations only known at runtime





DrillBeyond Operator ω



INIT()

Inititializes state

function INIT $state \leftarrow `collecting'$ $tuplestore \leftarrow \emptyset$ $auqMap \leftarrow HashMap()$

$n \leftarrow 0$

NEXT()

- Produces augmented tupels in three phase
 - Collect()
 - Augment()
 - Project()

```
function NEXT

if state = 'collecting' then

COLLECT()

AUGMENT()

state ← 'projecting'
return PROJECT()
```

DrillBeyond Operator ω (2)



COLLECT()

- Pulls and stores all tuples
- Blocking

AUGMENT()

Pass all entries from the EAS into a hashtable

```
function Collect while true do t \leftarrow \text{Next}(childPlan) if t = NULL then break tuplestore \leftarrow t augKey \leftarrow \text{TextAttrs}(t) if augKey \not\in augMap then augMap[augKey] \leftarrow \emptyset
```

function Augment

```
augReq \leftarrow (\forall k \in augMap \mid augMap[k] = \emptyset)

for all augKey, [augValues...] \in SEND(augReq) do

augMap[augKey] \leftarrow [augValues...]
```



DrillBeyond Operator ω (3)



PROJECT()

 Produces output tuples by replaying the stored tuples and filling the augmentation attribute

RESCAN()

 Called when subtrees have to be reexecuted

NEXT VARIANT()

Produces the multi-variant query results

function Project

 $t \leftarrow \text{Next}(tuplestore)$ **if** t = NULL **then return** NULL $augKey \leftarrow \text{TEXTATTRS}(t)$ t[augAttr] = augMap[augKey][n] **return** t

function RESCAN

 $state \leftarrow `collecting' \\ tuplestore \leftarrow \emptyset$

function NextVariant

RESCAN(tuplestore) $n \leftarrow n + 1$



Why Blocking?



EXAMPLE

- Send each tuple to the augmentation system at is own (tupleat-a-time)
- Augmentation system may choose ds₁, ds₂ and ds₄

country	gdp
Russia	?
UK	?
USA	?

MORE CONSISTENT: DS₁ AND DS₃

FOR QUALITY REASONS THE DRILLBEYOND OPERATOR...

- ...needs to be a **blocking** operator
- Realized in the "collecting" state in function Next()
- Consumes tuples from underlying operators until these are exhausted
- Hands them over to the augmentation system

ds₁ American Countries ds₃ World GDPs

country	gdp m USD	country	gdp Mil. \$
USA	Χ	UK	Χ
Canada	Υ	Russia	Z

ds₂ European Economy ds₄ European Economy

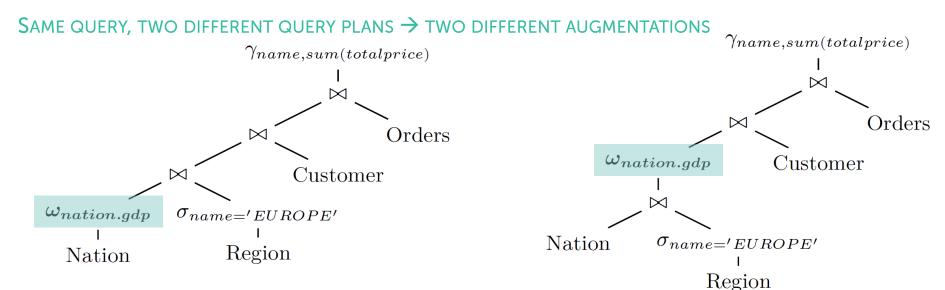
country	gdp EUR	
Russia	Χ	
France	Υ	

		-
country	gdp EUR	
UK	Χ	



Lower Placement Bound





LOWER PLACEMENT BOUND

- ullet ω can only be placed when all filters on R, e.g., joins and predicates, have been applied
- In other words: apply ω to the minimum number of entities in R

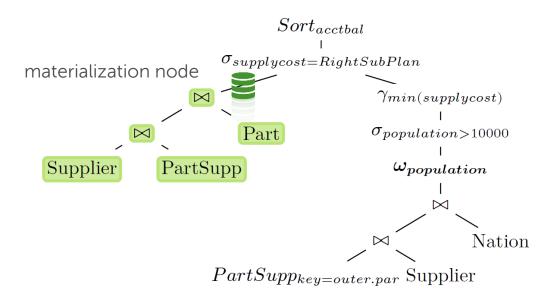


Invariant Caching



EXAMPLE

Modeled after TPCH query 2



LIMITATION: DEPENDING ON THE COST-BASED PLACEMENT OF ω , MOST OF THE QUERY MAY BE VARYING

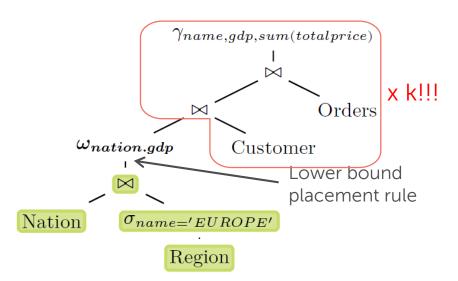


Maximize Invariant Sub-Trees



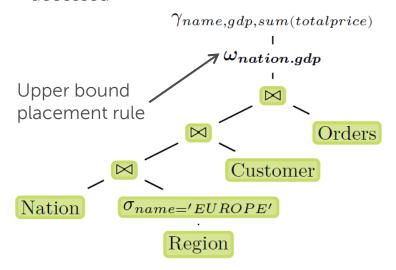
OPTIMAL PLAN...

...with respect to a single query execution



OPTIMAL PLAN...

- ...with respect to multiple query executions
- Upper bound placement rule: Place ω not earlier than the augmented values need to be accessed





Maximize Invariant Sub-Trees (2)

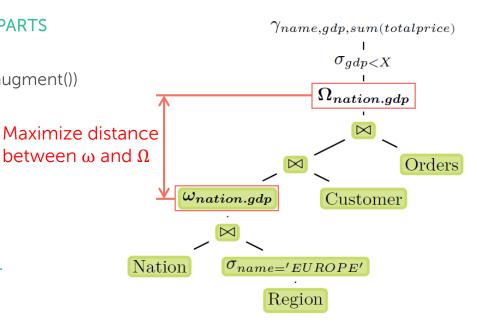


OBSERVATION

• We can separate the input part of ω from the actual projection of augmented values

SPLIT THE DRILLBEYOND OPERATOR INTO TWO PARTS

- ω
 - Performing the augmentation (collect() and augment())
 - Projects placeholders
 - Placed at the lower placement bound
- Ω
 - Performing the projection of values
 - Dereferences to the correct value array
 - Placed at the upper placement bound
- → MINIMIZING AUGMENTATION OPERATOR COST
- → MAXIMIZING SIZE OF INVARIANT QUERY PARTS



Dynamic Selection Pull-Up

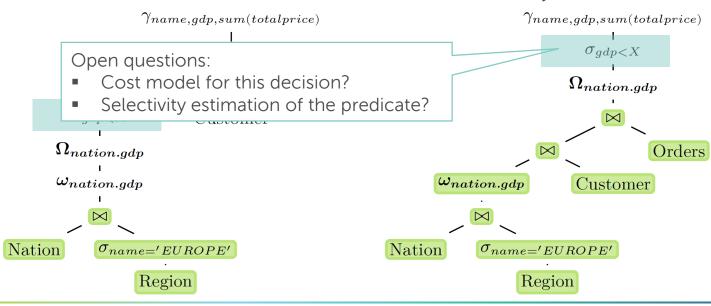


SELECTION PUSH-DOWN

- Smaller intermediate join result...
- ...but more varying nodes in the plan

SELECTION PULL-UP

- Place Ω as describe before
- Larger invariant subplan...
- ...but increased join results





Dynamic Selection Pull-Up (2)



COST MODEL

■ Binary decision \rightarrow decide for the minimum cost of the subtree above ω

$$C_{k}(\omega, \top) = \begin{cases} C_{1}(\omega, \Omega) + \sum_{i=1}^{k} s_{i} \cdot C_{1}(\omega, \top) & (Pull - up) \\ \sum_{i=1}^{k} s_{i} \cdot C_{1}(\omega, \top) & (no Pull - up) \end{cases}$$

- C(x,y): cost of the subplan from operator x to y
- S_i
 - Selectivity of the ith data source for the attribute in question
 - Not known at plan-time → determined in the augment() phase



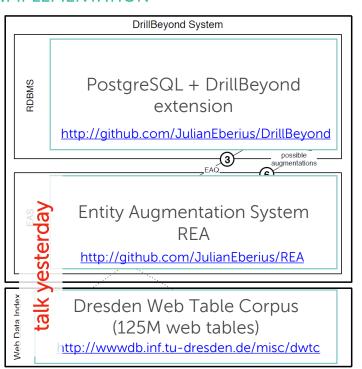
Evaluation



Experimental Setup



IMPLEMENTATION



TEST DATABASE

 Variation of TPC-H replacing generic identifiers with real-world entities, e.g. real company names for the Supplier relation

Test queries

- Subset of TPC-H queries in which dimension tables are used
- Added arbitrary where-clauses to one of the dimensions: "relation.X > Y"

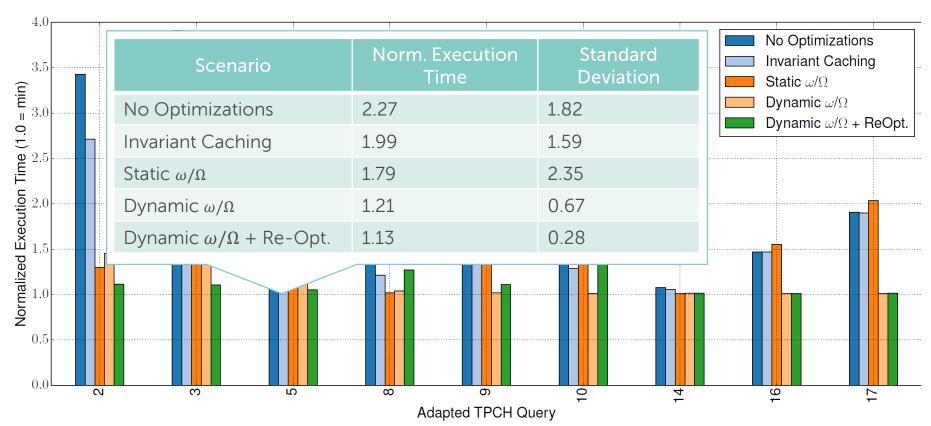
PARAMETERS

- Number of augmentations $k \in (1, 3, 5, 10)$
- Selectivity s ranging between 0.01 and 0.99 in ten steps



Overall Performance

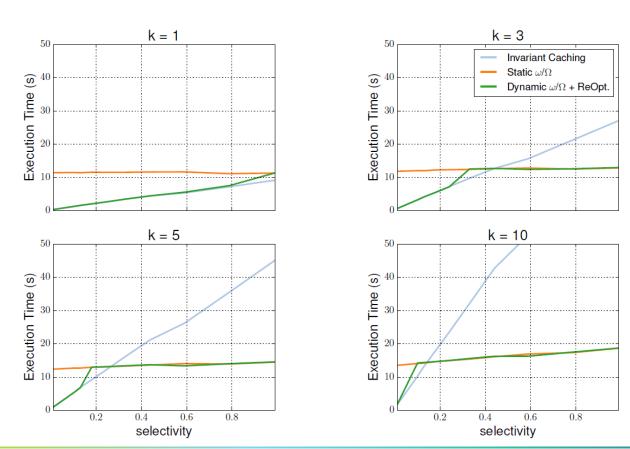






Performance of query 9, by selectivity







Summary



DRILLBEYOND A RDBMS / IR HYBRID SYSTEM

- Integration of top-k entity augmentation system into a RDBMS
- Multiple alternative query result solving the uncertainty and ambiguity of the automated integration

NEW OPERATORS: DRILLBEYOND ω/Ω

- Implemented in traditional interface functions: Init(), Next() and ReScan()
- Own cost model

NEW OPTIMIZATION STRATEGIES

- Invariant Caching
- Maximization of invariant sub-trees
- Selection push-down versus selection pull-up (runtime reoptimization)



