An Adaptive Query Execution Engine for Data Integration

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What makes Data Integration Systems (DIS) more challenging than traditional DB Systems?

- Query Reformulation
- The construction of wrapper programs
- Query optimizers and efficient query execution engines
Characteristics of DISs

- Unreliable or missing data statistics
- Unpredictable data transfer rates
- Unreliable, overlapping sources
- Want initial results quickly
- Network bandwidth generally constrains the data sources to be “small”

System needs to be adaptive
Tukwila Architecture

Data from sources

Execution Engine
- Event Handler
- Operator Execution

Temp Store

Execution Stats

Wrappers

Query

Reformulator

Logical Plan

Optimizer, Scheduler

Exec. Query Plan

Data Src. Cat.
Adaptivity through interleaving of planning and execution

Novel characteristics of Tukwila:

• The optimizer can create a partial plan if essential statistics are missing or uncertain
• The optimizer generates operator trees and appropriate event-condition-action rules
• Optimizer conserves the state of its search space when it calls the execution engine
• Dynamic collector operator to work with multiple sources
Discussion 1: Overlapping data sources

Tukwila handles data sources with overlaps, though the paper doesn't delve into the reasons for its significance. What challenges might arise due to overlapping data, and what are potential approaches to address them?

Discuss in groups of two.
Query Plan Execution

Query plan represented as data-flow tree:

- Control flow
  - **Iterator** (top-down)
    - Used by Tukwila
  - **Data-driven** (bottom-up)

“Show which orders have been delivered”

```
Select Status = "Delivered"
Join Orders.TrackNo = UPS.TrackNo
Read Orders
Read UPS
```
A **plan** includes a partially-ordered set of **fragments** and a set of **global rules**.

A **fragment** consists of a fully pipelined tree of **physical operators** and a set of **local rules**.

Key mechanism for adaptivity: At the end of each fragment, the rest of the plan can be re-optimized or rescheduled.

\[ \text{Select}_{\text{Status} = \text{"Delivered"}} \]
\[ \text{Read Orders} \]
\[ \text{Join}_{\text{Orders.TrackNo} = \text{UPS.TrackNo}} \]
\[ \text{Read UPS} \]
Implementation of Adaptivity: Rules

- **Reoptimization**
  - Reinvoke optimizer if cardinality estimates of results differ significantly

- **Contingent planning**
  - The execution engine checks properties of the result to select the next fragment

- **Rescheduling**
  - Reschedule if a source times out

- **Adaptive operators**
Tukwila Plans & Execution

When *event* if *condition* then *actions*

When *closed*(frag1)
  if *card*(join1)>2*est_card*(join1)  
  then replan

On event trigger, the rules are checked

Read Orders

Select _Status = “Delivered”_

Join _Orders.TrackNo = UPS.TrackNo_

When(closed(1)):
  if size_of(Orders) > 1000
  then reoptimize {2, 3}
Discussion 2: Tukwila’s motivations

Consider one of the following motivating situations of Tukwila:
1. Absence of statistics
2. Unpredictable data arrival characteristics
3. Overlap and redundancy among sources
4. Optimizing the time to initial answers

Answer these questions:
Q1: Can you give some examples where the chosen topic matters?
Q2: If you are a member of Tukwila team, what rules or policy would you have to deal with the problem?

Discuss in groups of 4, half system, half non-system.
Adaptive Query Operators
Why not conventional Joins?

Unpredictable Data Transfers
• Sort merge joins & indexed joins
  • Slow transfer blocks execution
• Nested loops joins and hash joins
  • Slow transfer of inner relationship, hard to choose inner relationship

Solution: Double Pipelined Hash Join
Double Pipelined Hash Join

- Proposed for parallel main-memory databases (Wilschut 1990)
  - One hash table per source
  - Add to hash table and probe opposite table as stream

Advantages:
- Data-driven: Results as soon as tuples are received
- Symmetric
- Can process faster data source while waiting for the slower source
Example

Orders

<table>
<thead>
<tr>
<th>OrderNo</th>
<th>TrackNo</th>
</tr>
</thead>
<tbody>
<tr>
<td>1234</td>
<td>01-23-45</td>
</tr>
<tr>
<td>1235</td>
<td>02-90-85</td>
</tr>
<tr>
<td>1399</td>
<td>02-90-85</td>
</tr>
<tr>
<td>......</td>
<td>......</td>
</tr>
</tbody>
</table>

UPS

<table>
<thead>
<tr>
<th>TrackNo</th>
<th>Status</th>
</tr>
</thead>
<tbody>
<tr>
<td>01-23-45</td>
<td>In Transit</td>
</tr>
<tr>
<td>02-90-85</td>
<td>Delivered</td>
</tr>
<tr>
<td>03-99-10</td>
<td>Delivered</td>
</tr>
<tr>
<td>......</td>
<td>......</td>
</tr>
</tbody>
</table>

Join \( \text{Orders.TrackNo} = \text{UPS.TrackNo} \) (Orders, UPS)

Hash Table (Orders)

01-23-45

Add

Hash Table (UPS)

No match!
## Example

### Orders

<table>
<thead>
<tr>
<th>OrderNo</th>
<th>TrackNo</th>
</tr>
</thead>
<tbody>
<tr>
<td>1234</td>
<td>01-23-45</td>
</tr>
<tr>
<td>1235</td>
<td>02-90-85</td>
</tr>
<tr>
<td>1399</td>
<td>02-90-85</td>
</tr>
<tr>
<td>......</td>
<td>......</td>
</tr>
</tbody>
</table>

### UPS

<table>
<thead>
<tr>
<th>TrackNo</th>
<th>Status</th>
</tr>
</thead>
<tbody>
<tr>
<td>01-23-45</td>
<td>In Transit</td>
</tr>
<tr>
<td>02-90-85</td>
<td>Delivered</td>
</tr>
<tr>
<td>03-99-10</td>
<td>Delivered</td>
</tr>
<tr>
<td>......</td>
<td>......</td>
</tr>
</tbody>
</table>

The join operation is formalized as:

$$\text{Join}_{\text{Orders.TrackNo} = \text{UPS.TrackNo}} (\text{Orders}, \text{UPS})$$

### Hash Tables

- **Hash Table (Orders)**
  - |TrackNo|
  - |01-23-45|
- **Hash Table (UPS)**
  - |TrackNo|
  - |01-23-45|

**Example:**

**Probe:** 01-23-45

**Output Match:** (01-23-45, 1234, In Transit)

**Add:**

<table>
<thead>
<tr>
<th>OrderNo</th>
<th>TrackNo</th>
<th>Status</th>
</tr>
</thead>
<tbody>
<tr>
<td>1234</td>
<td>01-23-45</td>
<td>In Transit</td>
</tr>
</tbody>
</table>

**Join**

$$\text{Join}_{\text{Orders.TrackNo} = \text{UPS.TrackNo}} (\text{Orders}, \text{UPS})$$
Memory Overflow

- **Memory heavy**: May not be able to fit both hash tables in RAM

- Strategies
  - **Incremental Left Flush**: Slowly degrade to hybrid hash join by flushing buckets of one source
  - **Incremental Symmetrical Flush**: Flush corresponding buckets of both sources
  - Set strategy through rules
Discussion 3: Tukwila

Would the adaptive behaviour of Tukwila be beneficial in general database systems? Would it boost efficiency?

What could be some advantages and disadvantages of applying the same methods to general database systems?

Discuss in groups of 2.
Summary

- Modified query optimizer to account for the unique characteristics of DISs
- Adaptivity through interleaving of optimization and execution
- Event-condition-action rule system
- Adaptive double pipelined hash join and collector