### Bridging the Archipelago between Row-Stores and Column-Stores for Hybrid Workloads

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### BACKGROUND (getting everyone on board with jargon)

- DBMS –
- OLTP –
- OLAP -
- HTAP –

- DBMS DataBase Management System
- OLTP –
- OLAP –
- HTAP –

- DBMS DataBase Management System
- OLTP On-Line Transaction Processing
- OLAP –
- HTAP –

- DBMS DataBase Management System
- OLTP On-Line **Transaction** Processing (HOT)
- OLAP –
- HTAP –

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- OLAP On-Line Analytical Processing
- HTAP –

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- OLAP On-Line Analytical Processing (COLD)
- HTAP Hybrid Transactional-Analytical Processing Workloads

- OLTP On-Line **Transaction** Processing (HOT)
- OLAP On-Line Analytical Processing (COLD)
- HTAP Hybrid Transactional-Analytical Processing
- NSM –
- DSM –

- OLTP On-Line **Transaction** Processing (HOT)
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- HTAP Hybrid Transactional-Analytical Processing
- NSM *n*-ary Storage Model
- DSM –

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- HTAP Hybrid Transactional-Analytical Processing
- NSM *n*-ary Storage Model (Why is this good for writes?)
- DSM –

- OLTP On-Line **Transaction** Processing (HOT)
- OLAP On-Line Analytical Processing (COLD)
- HTAP Hybrid Transactional-Analytical Processing
- NSM *n*-ary Storage Model (Why is this good for writes?)
- DSM Decomposed Storage Model

Storage models

### Here's your pneumonic device



#### OLTP



OLAP

Tea is hot

**new** data (Updates and writes)

n-ary Storage

You Analyze history

History is (c)old (Scans and aggregations)

**Decomposed** Storage

THE PROBLEM

### Today we're concerned with HTAP

- Not just a "dynamic workload"
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- Not just a "dynamic workload"
- Transactions and analytics queries running simultaneously
- Both historical and fresh data are equally relevant to analysis
- Examples?

### You might work with IoT sensors



### Or you might run a search engine



# Or you might feel positively about the concept of money



# One approach is to physically separate the use cases

System A gives data to System B at some point!	System A	System B	
Workload	OLTP	OLAP	
Storage Model	n-ary	Decomposed	
Data stored as	Rows	Columns	
Used for	Inserts & Updates	Reads	

### Is this really HTAP?

System A gives data to System B at some point!	System A	System B	
Workload	OLTP	OLAP	
Storage Model	n-ary	Decomposed	
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# Separating the use cases defeats the purpose of HTAP



• What are they?

- What are they?
  - Two different execution engines

- What are they?
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    - Twice the software

- What are they?
  - Two different execution engines
    - Twice the software
      - Twice the people!
        - (at least twice the cost...)

THE SOLUTION

# A flexible storage model (FSM) takes the "temperature" of tuples



Source: D. Beaver, S. Kumar, H. C. Li, J. Sobel, P. Vajgel, and F. Inc. *Finding a needle in haystack: Facebook's photo storage*. In OSDI, 2010.

# A flexible storage model (FSM) takes the "temperature" of tuples

ID	IMAGE-ID	NAME	PRICE	DATA
101	201	ITEM-101	10	DATA-101
102	202	ITEM-102	20	DATA-102
103	203	ITEM-103	30	DATA-103
104	204	ITEM-104	40	DATA-104

(a) OLTP-oriented N-ary Storage Model (NSM)

ID	IMAGE-ID	NAME	PRICE	DATA
101	201	ΠΕΜ-101	10	DATA-101
102	202	ITEM-102	20	DATA-102
103	203	ITEM-103	30	DATA-103
104	204	ITEM-104	40	DATA-104

(b) OLAP-oriented Decomposition Storage Model (DSM)

ID	IMAGE-ID	NAME	PRICE	DATA
101	201	ITEM-101	10	DATA-101
102	202	ITEM-102	20	DATA-102
103	203	ITEM-103	30	DATA-103
104	204	ITEM-104	40	DATA-104

(c) HTAP-oriented Flexible Storage Model (FSM)

### A "tile" is part row, part column

	ID	IMAGE-ID	NAME	PRICE	DATA	
ĺ	Tile 101	201	ITEM-101	Tile 10	DATA-101	File
	A-1 102	202	ITEM-102	A-2 20	DATA-102	Group
	103	203	ITEM-103	30	DATA-103	A
ſ	Tile 104	204 Tile	ITEM-104	40 Tile	DATA-104	Tile
	B-1 105	205 B-2	ITEM-105	50 B-3	DATA-105	Group
	106	206	ITEM-106	60	DATA-106	B
ſ	Tile 107 C-1 108 109 110	207 208 209 210	ITEM-107 ITEM-108 ITEM-109 ITEM-110	70 80 90 100	DATA-107 DATA-108 DATA-109 DATA-110	File Group C

Figure 3: Physical Tile – An example storage layout of a table composed of physical tiles. This table comprises of three tile groups (A, B, C).



Figure 4: Logical Tile – An example of a logical tile representing data spread across a couple of physical tiles (A-1, A-2).

### "Physical tiles" store subsets of tuple attributes

	ID	IMAGE-ID	NAME	PRICE	DATA	
Tile A-1	101 102 103	201 202 203	ПЕМ-101 ПЕМ-102 ПЕМ-103	Tile 10 A-2 20 30	DATA-101 DATA-102 DATA-103	Tile Group A
Tile B-1	104 105 106	204 Tile 205 B-2 206	ITEM- 104 ITEM- 105 ITEM- 106	40 Tile 50 B-3 60	DATA-104 DATA-105 DATA-106	Tile Group B
Tile C-1	107 108 109 110	207 208 209 210	ПТЕМ-107 ПТЕМ-108 ПТЕМ-109 ПТЕМ-110	70 80 90 100	DATA-107 DATA-108 DATA-109 DATA-110	Tile Group C

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### Oh, we can't?

# Why can't we just put physical tiles in our favorite DBMS?

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• Two words:

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# Query execution!

# "Logical tiles" store information about multiple physical tiles



# Logical tile columns contain sets of physical tiles columns

 The underlying physical data are released during materialization

# "Tile algebra" is an abstracted extension of relational algebra



### Tile algebra offers several advantages

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• Single execution engine

Vectorized processing (tiles instead of tuples)

 DBMS can optimize what materializes when and what goes in the cache

# The paper goes into detail about concurrency protocols

• And if you care about that, I invite you to read the paper!

• How might we do this?

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  - Copy data to optimal layout before executing query?

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• Background process reorganize one tile at a time?

- How might we do this?
  - Copy data to optimal layout before executing query?



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  - ... is that a problem?

Record attributes found in SELECT and WHERE clauses

• Only do this for a random sample of queries

- We have millions of writes and only a few big reads...
  - ... is that a problem?
- Record the cost of the queries too

# These statistics are used to repartition the tables into new tiles

<u>Clustering algorithm</u> chooses which attributes belong together in physical tiles

 <u>Greedy algorithm</u> groups tiles together based on how "important" they are to workloads

This is done incrementally to <u>amortize</u> the cost

### EVALUATION (NSM vs. DSM vs. FSM)

# The system was evaluated using workloads based on these queries

- $Q_1$ : INSERT INTO R VALUES  $(a_0, a_1, \ldots, a_p)$
- $Q_2$ : SELECT  $a_1, a_2, \ldots, a_k$  FROM R WHERE  $a_0 < \delta$
- $Q_3$ : SELECT MAX $(a_1), \ldots, MAX(a_k)$  FROM R WHERE  $a_0 < \delta$
- $Q_4$ : SELECT  $a_1 + a_2 + \ldots + a_k$  FROM R WHERE  $a_0 < \delta$
- $Q_5$ : SELECT X. $a_1$ ,...,X. $a_k$ ,Y. $a_1$ ,...,Y. $a_k$

FROM R AS X, R AS Y WHERE  $X.a_i < Y.a_j$ 

Note that different values for k and  $\delta$  alter the projectivity and the selectivity of the queries, respectively. We use different workloads comprised of these query types to evaluate the impact of the storage models on the performance of the DBMS.

# Narrow => 50 attributes; Hybrid => 1M writes per read



# Wide => 500 attributes; Aggregate => MAX(x, y, z, ...)



# We can see how FSM learns over time



# CONCLUSION

# I think the paper did a pretty good job of...

Demonstrating the importance of the problem and their solution

• (Usually) going into the right amount of detail

• Talking about different ways to implement each step

• Conducting a lot of difference experiments

### I wish the paper had...

• Done more join queries besides a few self-joins (major)

 Said either way more or slightly less about tiles (I know they said way more in the appendix, but... yikes)

 Benchmarked their approach against another HTAP solution (there was some hand waving in their critiques of such systems)