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

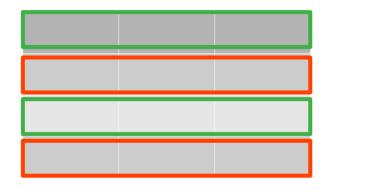
J Arulraj, et el.

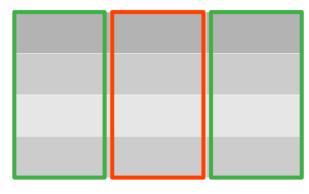
SIGMOD '20

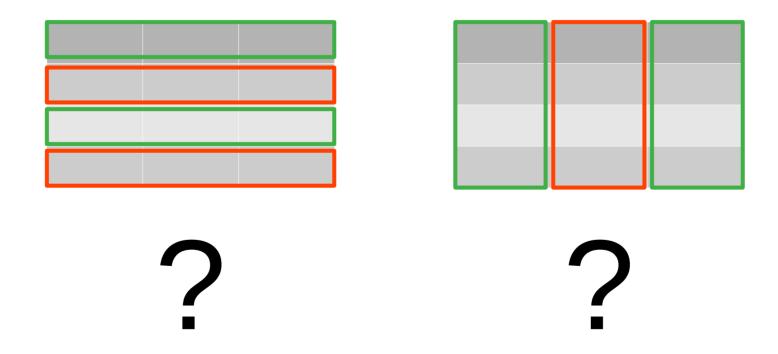
Speaker: Yu-Cheng Huang

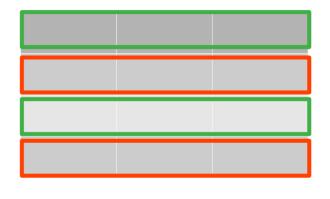
Introduction

What are the two basic memory layout?



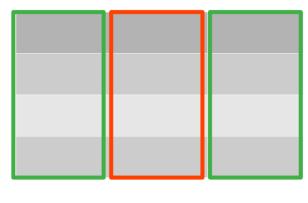






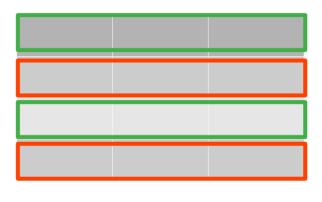
tuple-centric

Row Storage



Column Storage

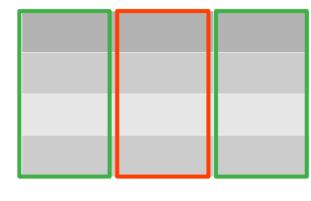
Columnar



Row Storage

tuple-centric

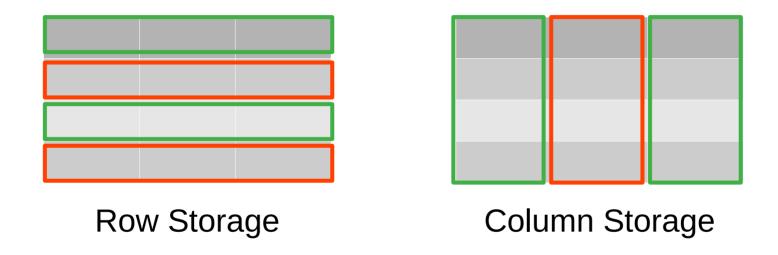
NSM



Column Storage

Columnar

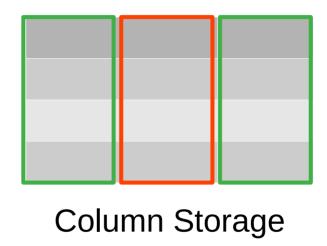
DSM



When?



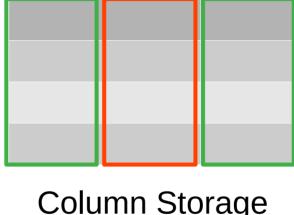




Search for an attribute



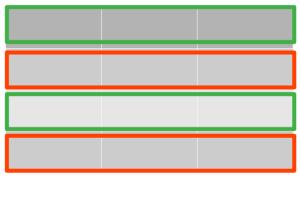
Row Storage



Column Storage

When? Insert, Update Search for an attribute

Data?

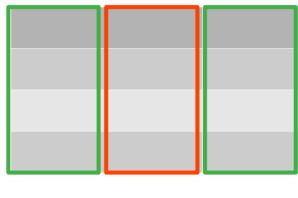


Row Storage



Data? Hotly updated

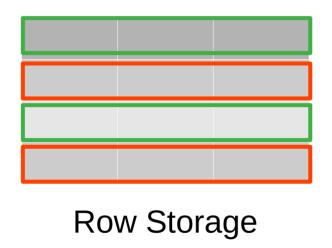
When?

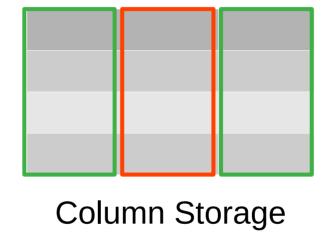


Column Storage

Search for an attribute

Coldly stored

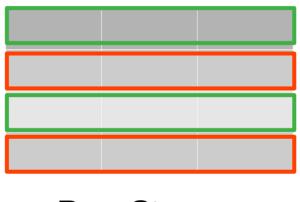




When? Insert, Update

Search for an attribute

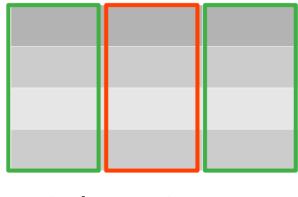
process?



Row Storage

When? Insert, Update

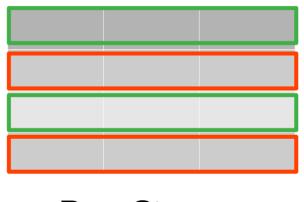
process? Transactional



Column Storage

Search for an attribute

Analytical



Row Storage



Column Storage

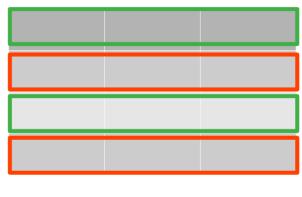
When? Insert, Update

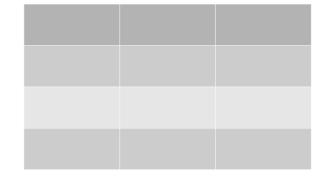
OLTP

Search for an attribute

OLAP

process?





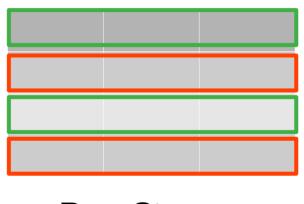
Row Storage

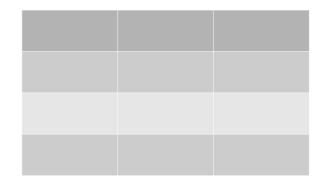
When?

Insert, Update

Search for an attribute

OLTP





Row Storage

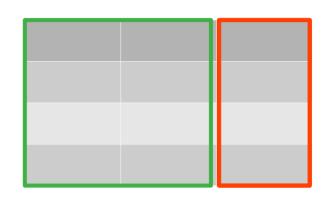
When?

Insert, Update

Search for (a1,a2) attribute

OLTP





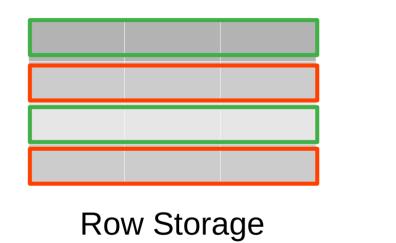
Row Storage

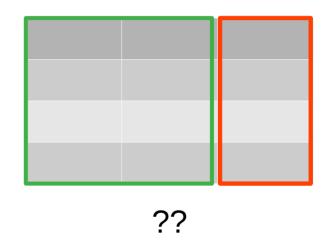
When?

Insert, Update

Search for (a1,a2) attribute

OLTP

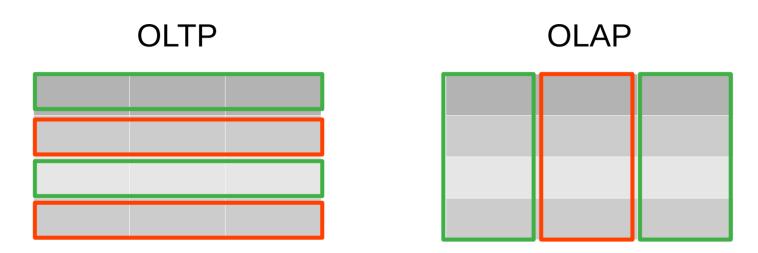


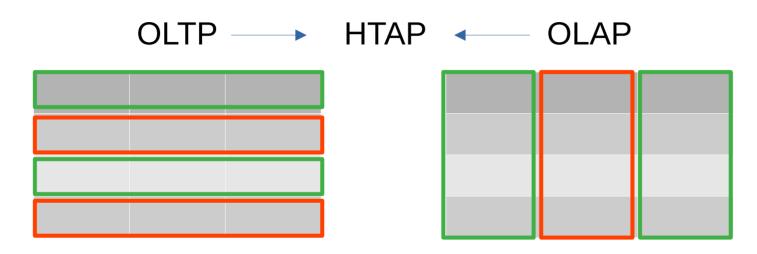


When? Insert, Update

Search for (a1,a2) attribute

OLTF

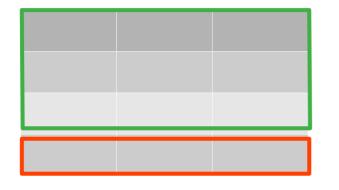


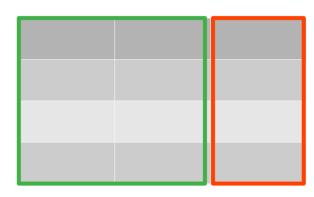


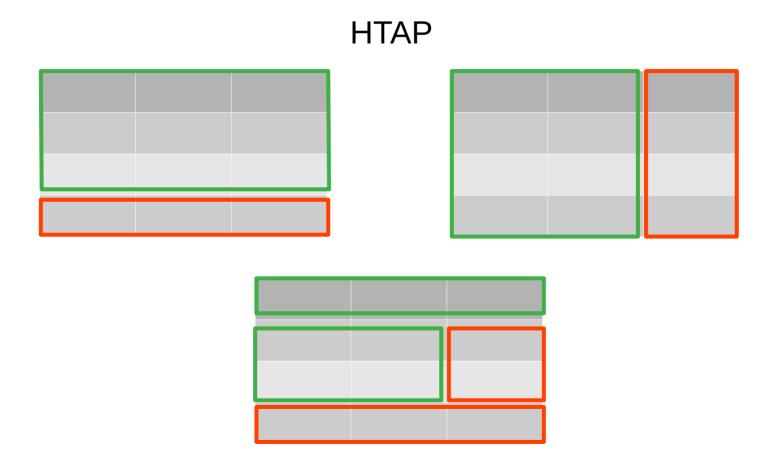
OLTP --- OLAP

HTAP

HTAP







Which?

Which?

Do we have any observation?

Which?

Do we have any observation?

Hot data → row Cold data → column

Hot data → row Cold data → column

To do such data distribution, what should be done?

Record Query types?

Record Query types?

If we know what types of queries there are, then we definitely can make a good design.

Record Query types?

If we know what types of queries there are, then we definitely can make a good design.

But ...

Record Query types?

If we know what types of queries there are, then we definitely can make a good design.

But ...

Can we be smarter?

Record Query types?

Self-adaptive?

Self-adaptive algorithm



Record

Self-adaptive algorithm

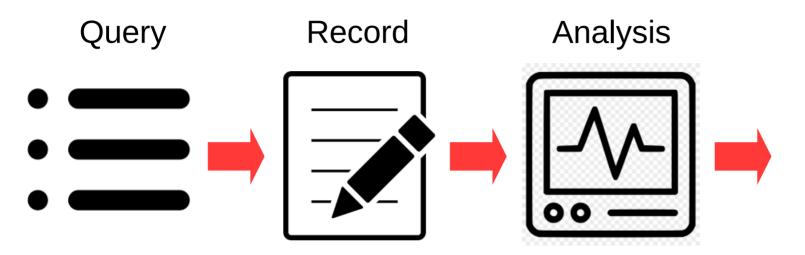


Record

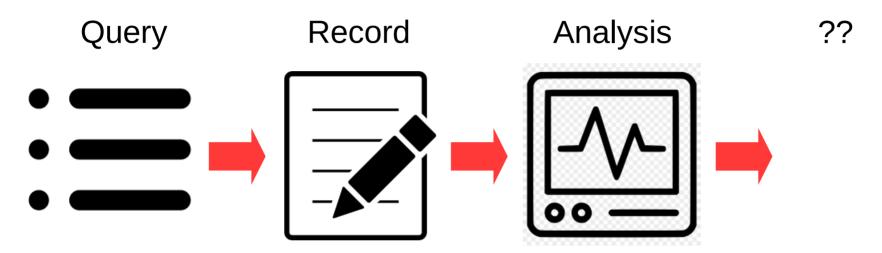


Analysis

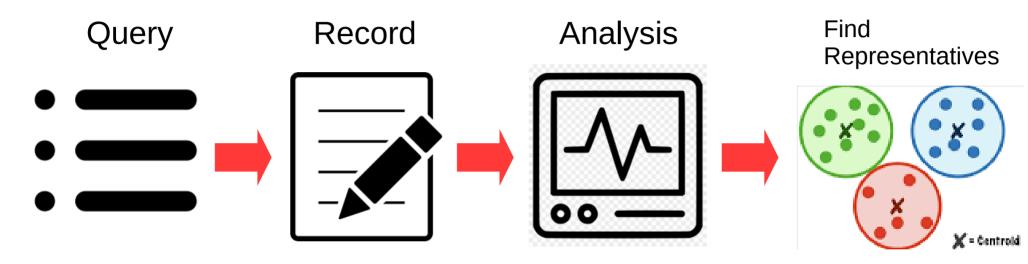
Self-adaptive algorithm



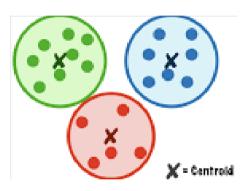
Self-adaptive algorithm



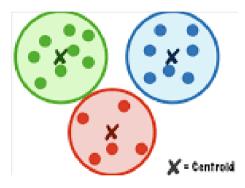
Self-adaptive algorithm



Find Representatives



Find Representatives



1st step:

Using recent n Queries,

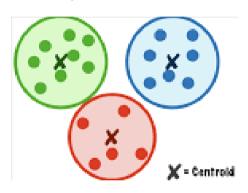
Find k Representatives R

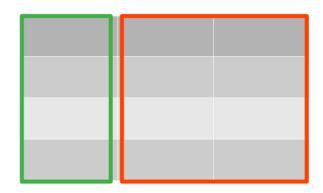
2nd step:

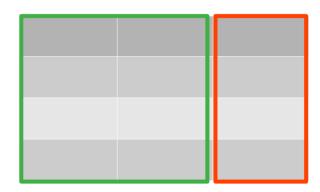
Generate vertical partitioned layout using R with greedy algorithm. (Largest cluster first)

Find Representatives

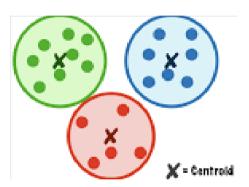
FSM (Flexible Storage Model)

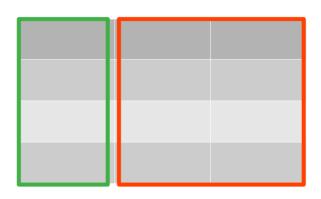


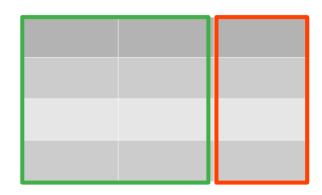




Find Representatives FSM (Flexible Storage Model)

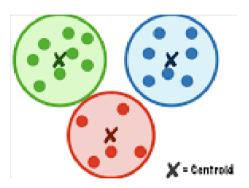


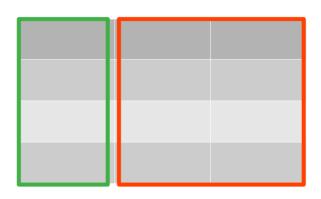


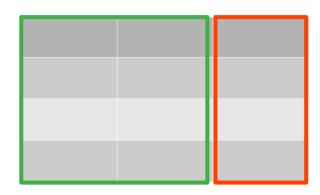


Different tables, different vertical layouts

Find Representatives FSM (Flexible Storage Model)

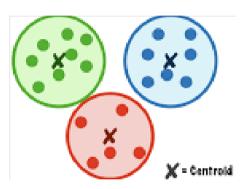


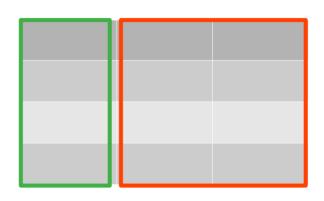


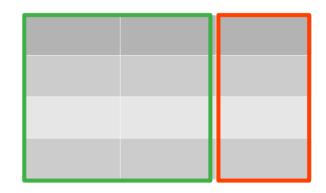


Different layouts → different access methods?

Find Representatives FSM (Flexible Storage Model)





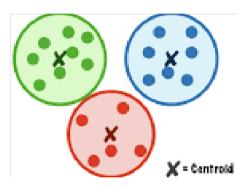


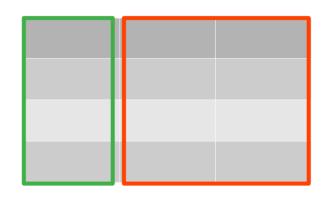
Different layouts → different access methods?

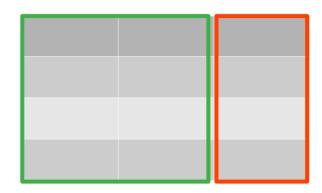
Inefficient!!



Find Representatives FSM (Flexible Storage Model)

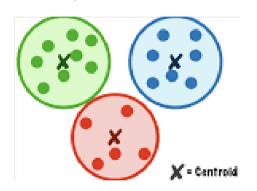


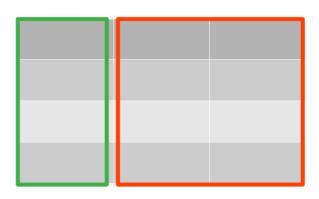


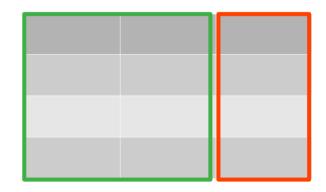


Provide Abstract layer!!

Find Representatives FSM (Flexible Storage Model)







Provide Abstract layer!!



One mutual access interface.

```
SELECT R.c, SUM(S.z)
FROM R JOIN S ON R.b = S.y
WHERE R.a = 1 AND S.x = 2
GROUP BY R.c;
```

```
SELECT R.c, SUM(S.z)
                 FROM R JOIN S ON R.b = S.y
                WHERE R.a = 1 AND S.x = 2
                GROUP BY R.c;
               Ω
                                        Materialize, { LT }, { PT }
          \Gamma_{c;sum(z)}
                                       Aggregate, { LT, C}, { LT }
          \bowtie_{\mathcal{R}.b=\mathcal{S}.y}
                                          Join, { LT, LT }, { LT }
                         \pi_{y,z}
                                         Projection, { LT }, { LT }
  \pi_{b,c}
                                     Sequential Scan, { T, P}, { LT }
\sigma_{R,a=1}
                       \sigma_{S,x=2}
   R
                                                   Table
```

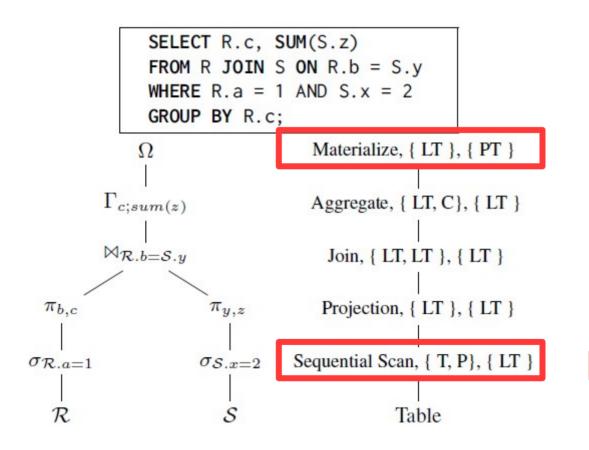
LT: logical tile

PT: physical tile

T: table

Attributes: C

Predicate: P



LT: logical tile

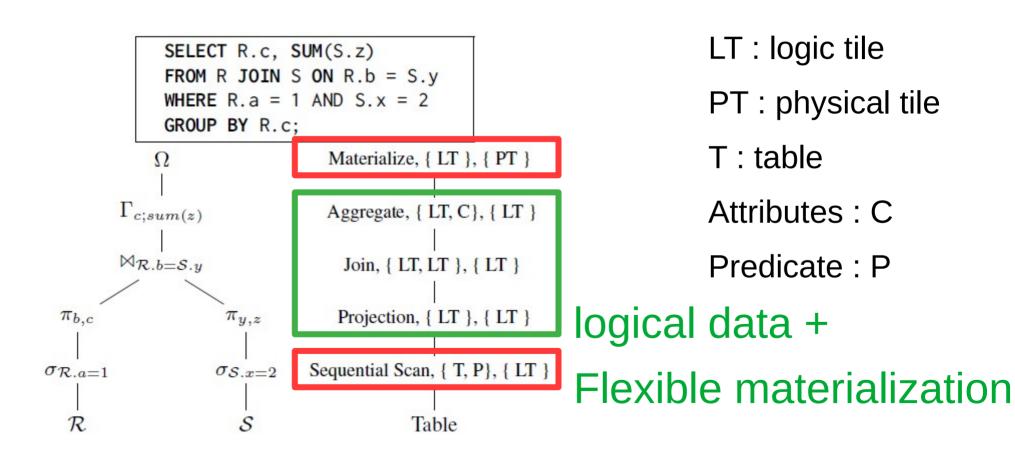
PT: physical tile

T: table

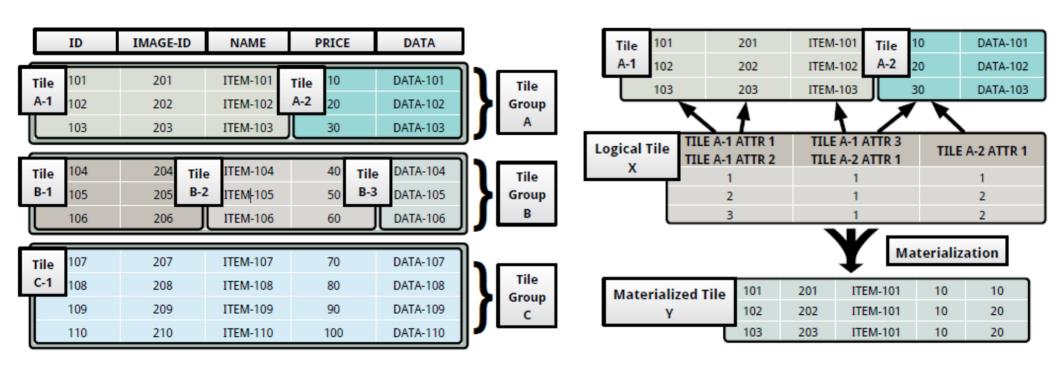
Attributes: C

Predicate: P

Physical data



Tiles



Physical Tile

Logical Tile

Definitely we do not want stalling.

Definitely we do not want stalling.

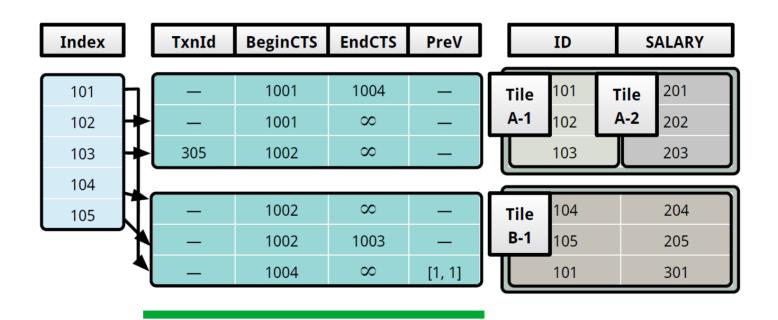
Write while reconfigure memory layout

Definitely we do not want stalling.

Write while reconfigure memory layout

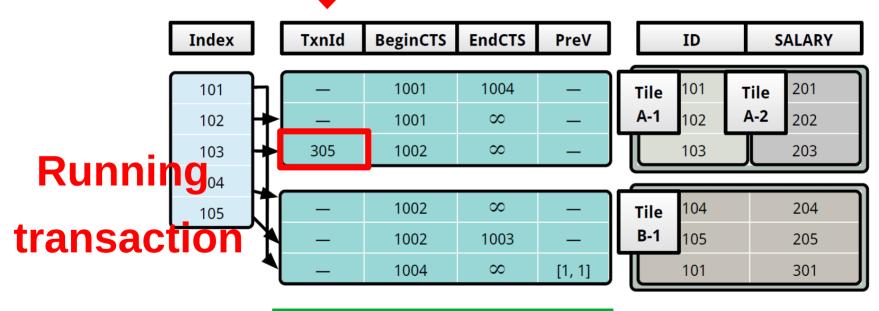
 \rightarrow MVCC

(Multi-Version Concurreny Control)

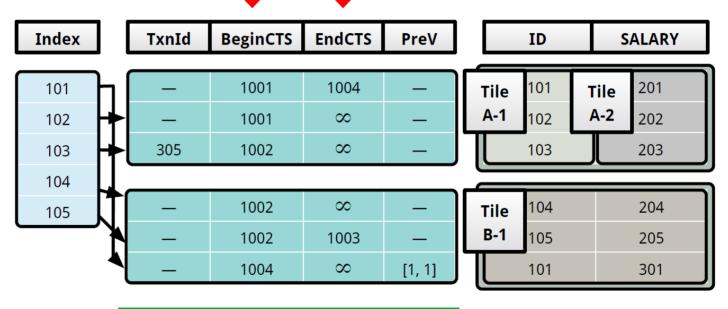


MVCC BeginCTS EndCTS Index TxnId PreV ID **SALARY** 1001 1004 201 101 Tile Tile A-1 A-2 102 202 102 1001 ∞ 103 305 1002 ∞ 203 103 104 1002 ∞ 204 Tile 105 B-1 1003 105 205 1002 101 1004 ∞ [1, 1] 301

MVCC Unique transaction identifier

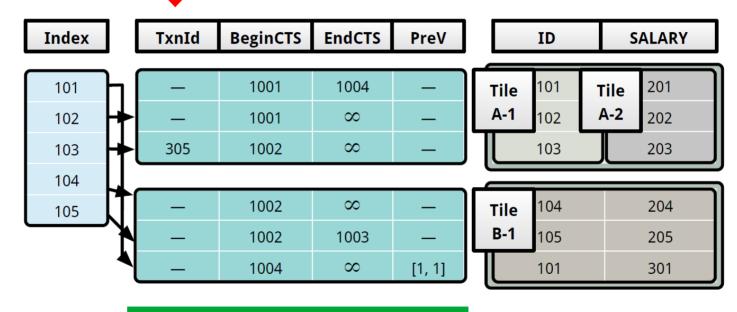


MVCC Unique commit timestamp



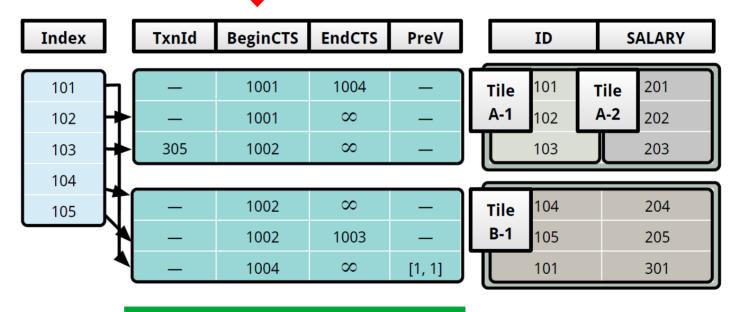
i) Insert

(atomic)



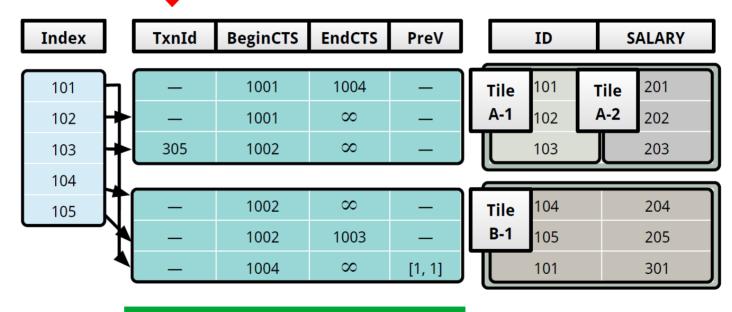
i) Insert

(after commit)

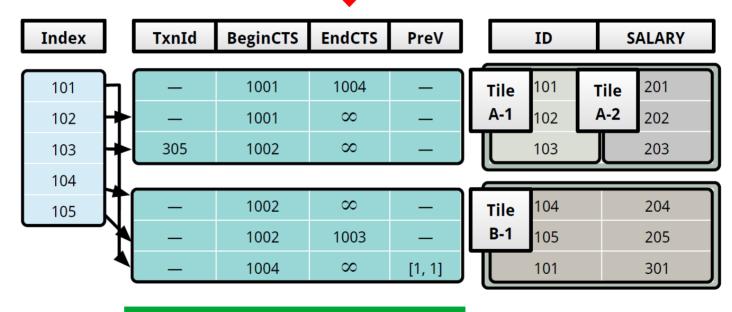


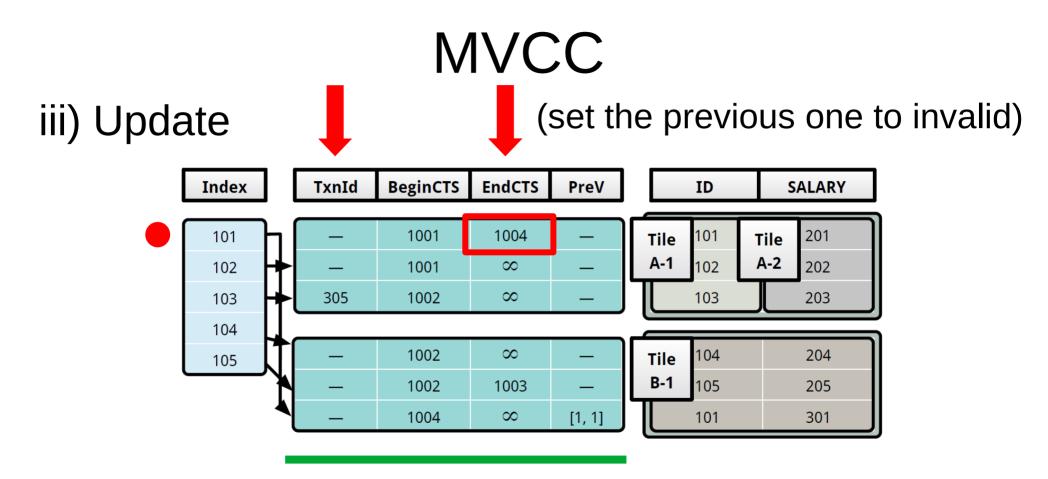
ii) Delete

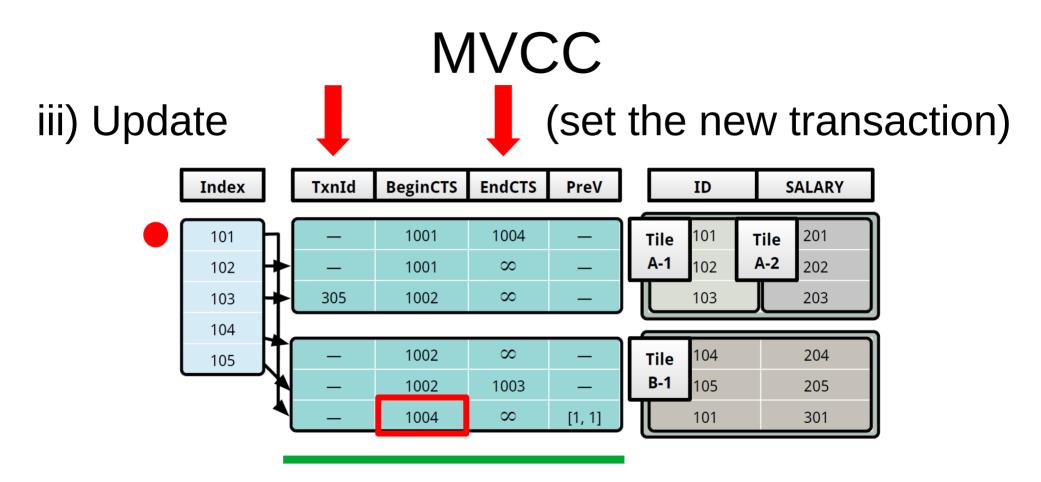
(atomic)

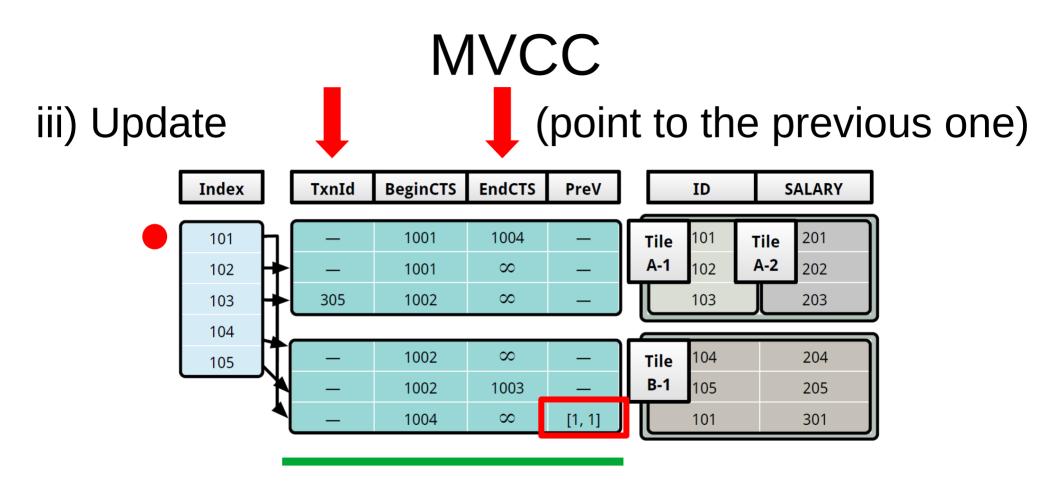


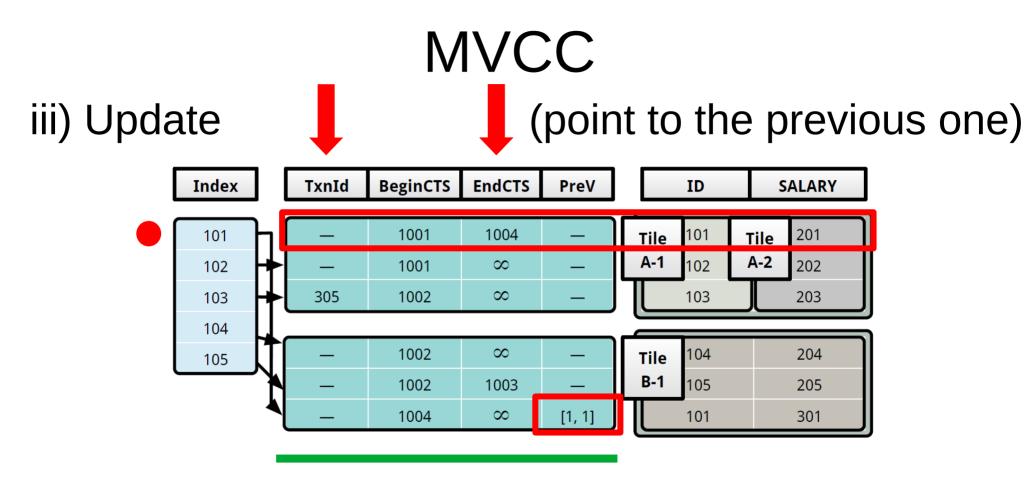
ii) Delete MVCC (after commit)

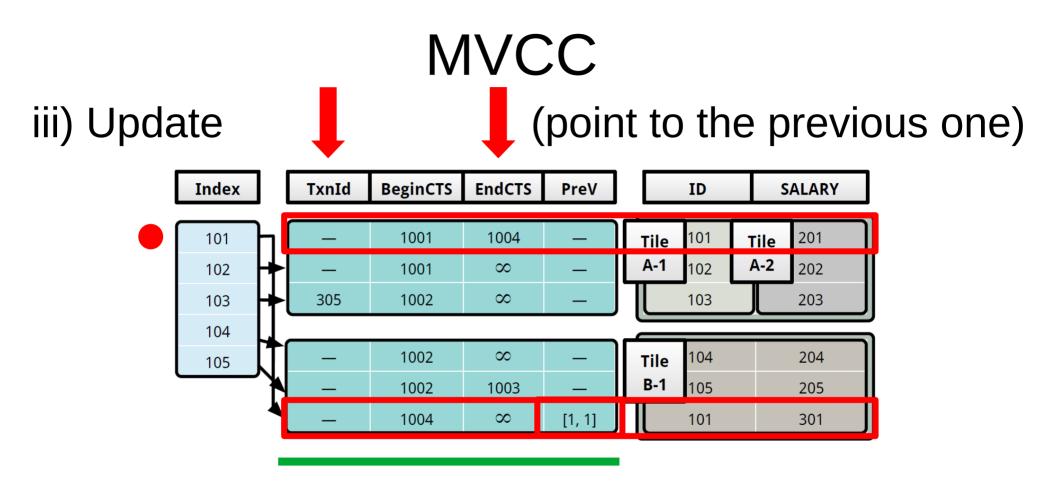


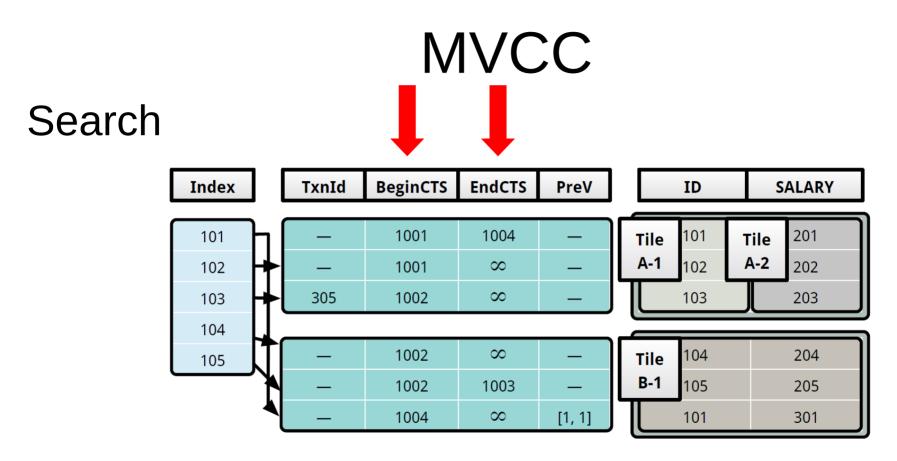




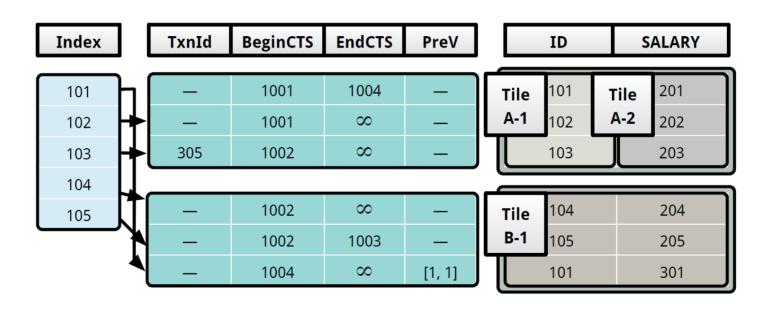








Looking through the metadata, find the one within Its visibility (BeginCTS < transaction ID < EndCTS)



Garbage Collector claims those old spaces back

Performance

What about the overhead?

Is this method practical?

Performance

What about the overhead?

Is this method practical?

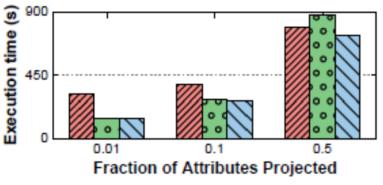
Let us see the results.

Evaluation

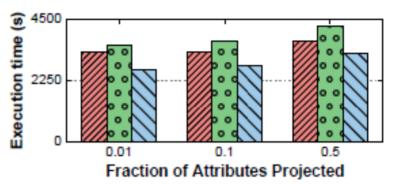
Experiemnt settings

- CPU: dual-socket Intel Xeon E5-4620 server
- OP: Ubuntu 14.04 (64-bit)
- Core: 2.6 GHz * 8 / socket
- DRAM: 128 GB
- L3 cache size: 20 MB

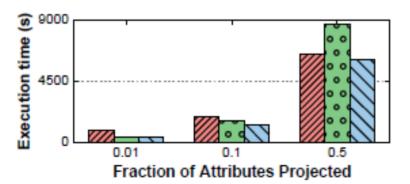
NSM / DSM / FSM SCAN



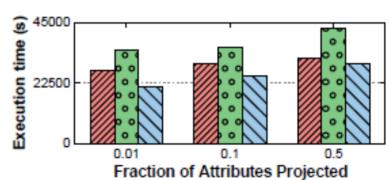
(a) Scan, Narrow, Read Only



(b) Scan, Narrow, Hybrid



(c) Scan, Wide, Read Only



(d) Scan, Wide, Hybrid

Storage Models:



NSM

00000



DSM

FSM

Adaptation











Conclusion

Conclusion

- Memory Layout (column storage, row storage)
- + Adaptation → FSM
- + Abstraction → Logical Tiles
- + Concurrency control → MVCC

Discussion

Discussion

 When the reorganization benefits the queries and whet it does not?

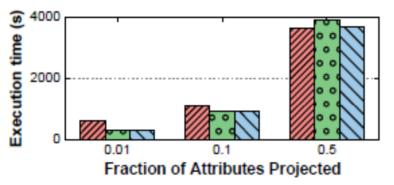
Discussion

- When the reorganization benefits the queries and whet it does not?
- Is k-means a good algorithm for the layout reorganization task?

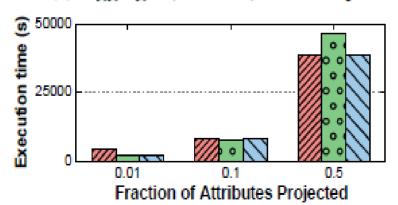
END

- 1 beginning, main contribution of the paper
- 2. before conclusion, whether it supports its idea
- 3. Conclude, open question
- 4.Experiement settings
- 5.mentioned NSM, DSM in the beginning part
 6. k means not really finding the right clusters (6)
 - 6. k-means \rightarrow not really finding the right clusters (CS565)
- 7. Asking questions

NSM / DSM / FSM Aggregate



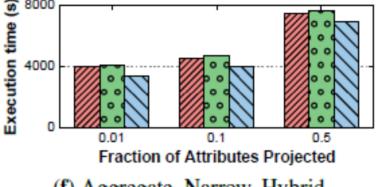
(e) Aggregate, Narrow, Read Only



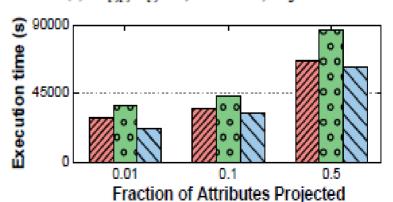
(g) Aggregate, Wide, Read Only

Storage Models :





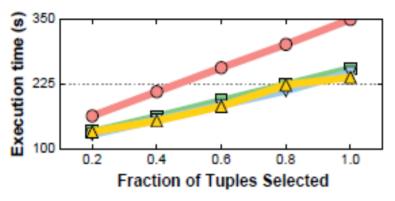
(f) Aggregate, Narrow, Hybrid



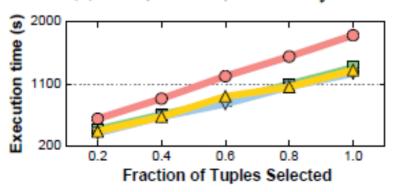
(h) Aggregate, Wide, Hybrid

|||||| FS

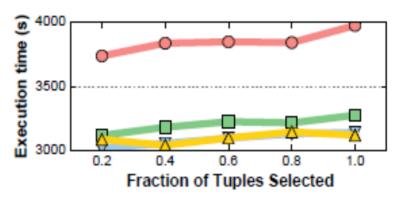
Horizontal Separation



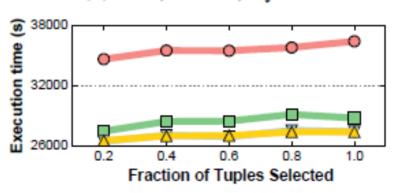
(a) Scan, Narrow, Read Only



(c) Scan, Wide, Read Only



(b) Scan, Narrow, Hybrid



(d) Scan, Wide, Hybrid

Storage Models:



NSM

00000

DSM



