UpBit: Scalable In-Memory Updatable Bitmap Indexing

Manos Athanassoulis[1] Zheng Yan[2] Stratos Idreos[1]

[1]Harvard University [2]University of Maryland

HARVARD UNIVERSITY

NSF

FNSNF

Indexing for Analytical Workloads

Column A	A=10	A=20	A=30
30	0	0	1
20	0	1	0
30	0	0	1
10	1	0	0
20	0	1	0
10	1	0	0
30	0	0	1
20	0	1	0

Specialized indexing

- ✓ Compact representation of query result
- ✓ Query result is readily available

Bitvectors

- ✓ Can leverage fast Boolean operators
- ✓ Bitwise AND/OR/NOT faster than looping over meta data

Bitmap Indexing Limitations

Column A	A=10	A=20	A=30
30	0	0	1
20	0	1	0
30	0	0	1
10	1	0	0
20	0	1	0
10	1	0	0
30	0	0	1
20	0	1	0

Index Size

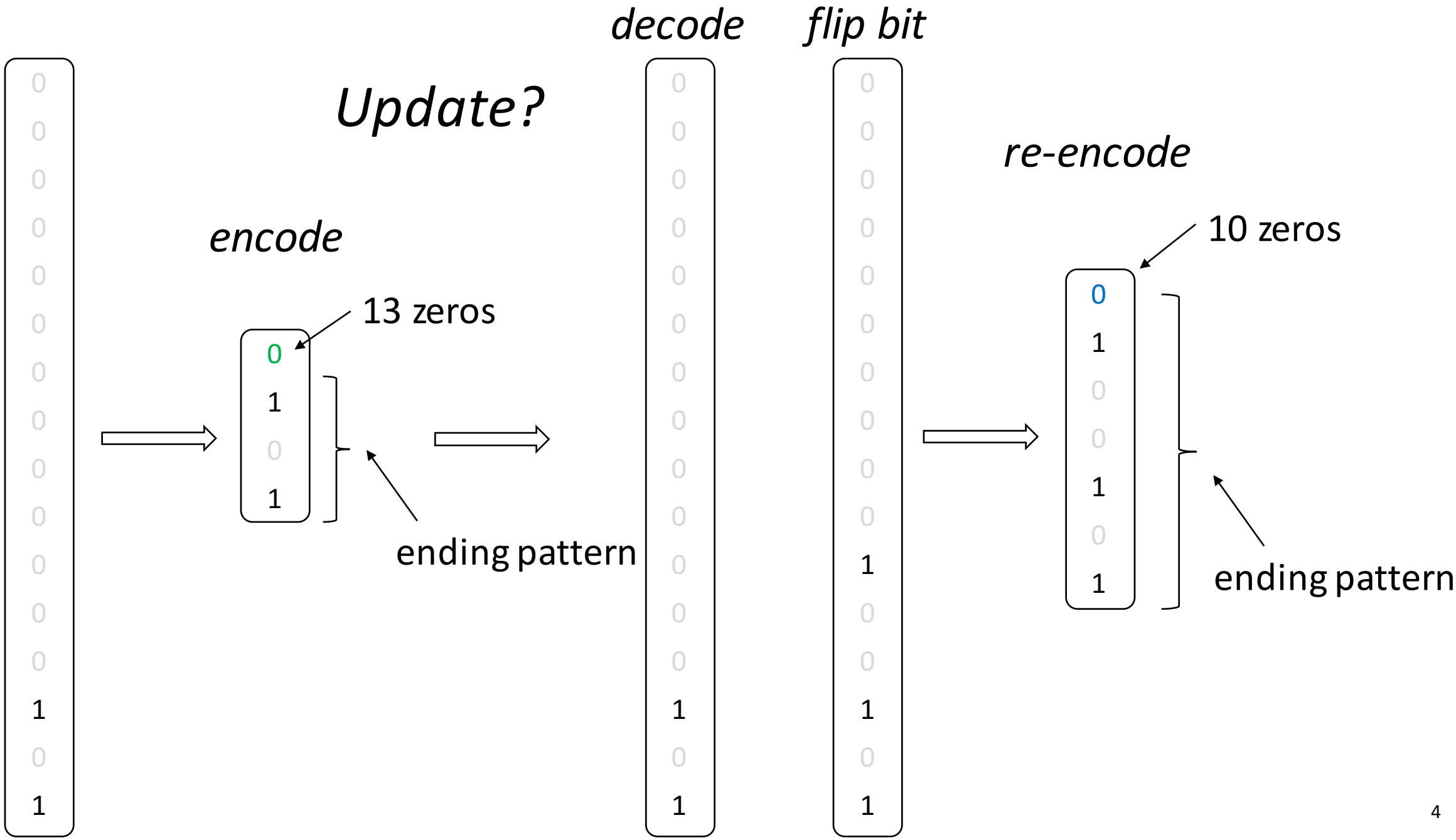
☒ Space-inefficient for large domains

☑ Addressed by bitvector encoding/compression

core idea: *run-length encoding* in prior work

but …

☒ Updating encoded bitvectors is **very** inefficient


Update?
encode
13 zeros
ending pattern
decode
flip bit
re-encode
10 zeros
ending pattern

Goal

Bitmap Indexing with efficient Reads & Updates

Prior Work: Bitmap Indexing and Deletes

Update Conscious Bitmaps (UCB), SSDBM 2007

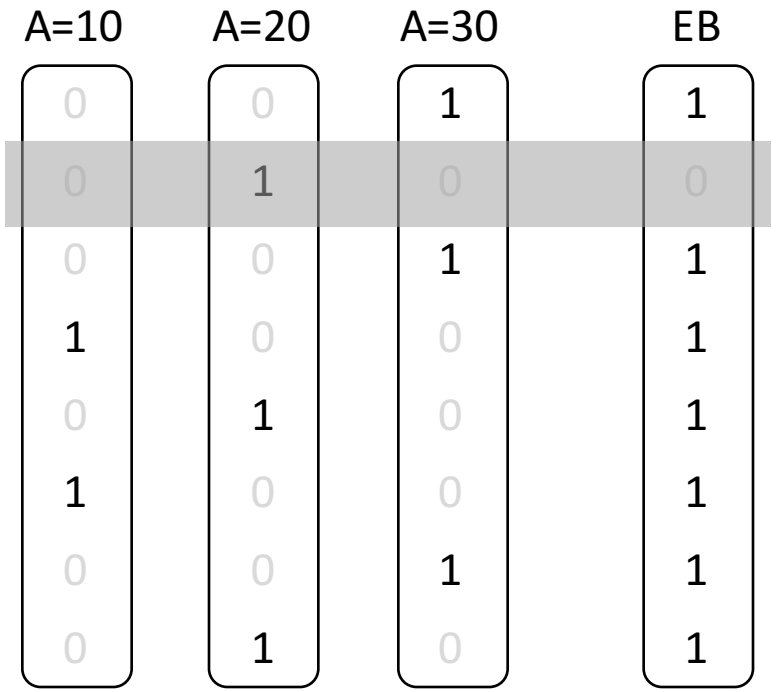
A=10	A=20	A=30	EB
0	0	1	1
0	1	0	1
0	0	1	1
1	0	0	1
0	1	0	1
1	0	0	1
0	0	1	1
0	1	0	1

efficient deletes by invalidation

existence bitvector (EB)

Prior Work: Bitmap Indexing and Deletes

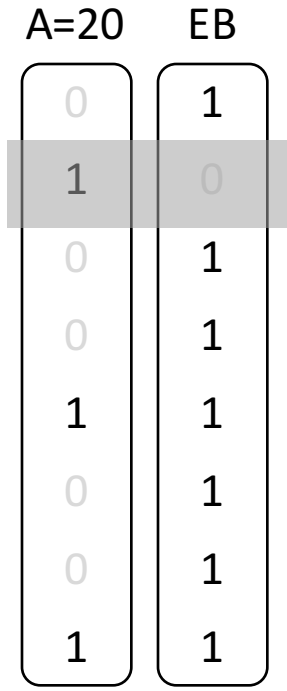
Update Conscious Bitmaps (UCB), SSDBM 2007



efficient deletes by invalidation
existence bitvector (EB)

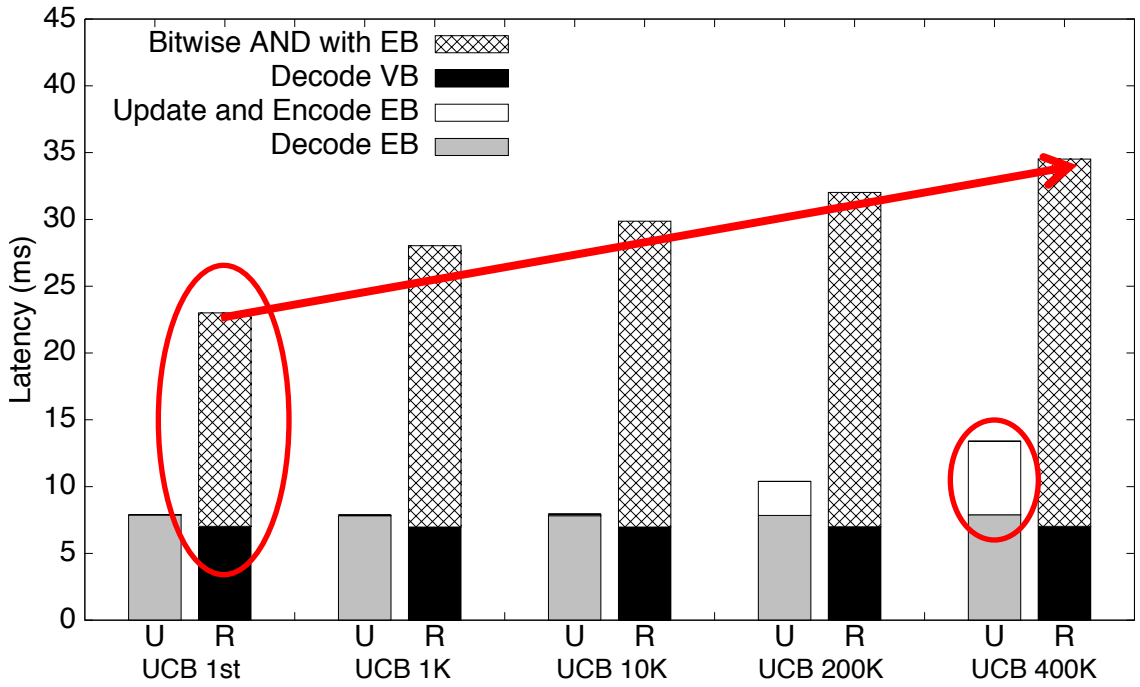
reads?
bitwise AND with EB

updates?
delete-then-append



Prior Work: Limitations

n=100M tuples, d=100 domain values, 50% updates / 50% reads



read cost increases with #updates

why?

bitwise AND with EB is the bottleneck

update EB is costly for >> #updates

UCB performance does not scale with #updates

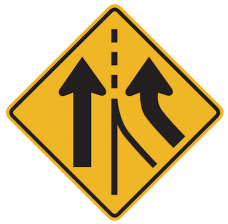
single auxiliary bitvector

repetitive bitwise operations

Bitmap Indexing for Reads & Updates

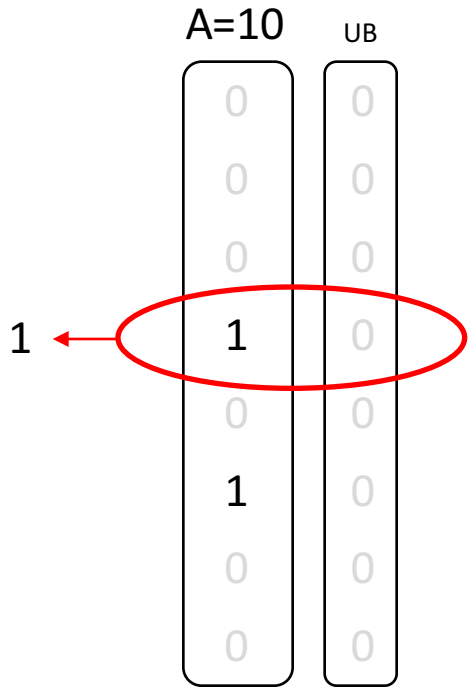
distribute update cost

efficient random accesses in compressed bitvectors



query-driven re-use results of bitwise operations

Design Element 1: update bitvectors

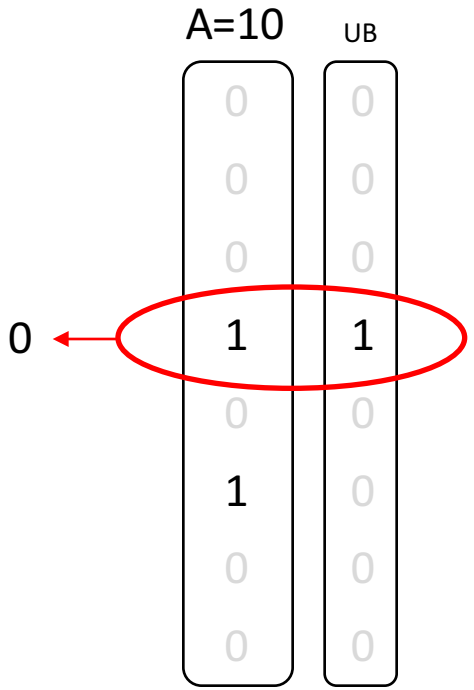


one per value of the domain
initialized to 0s

the current value is the XOR

every update flips a bit on UB

Design Element 1: update bitvectors



one per value of the domain
initialized to 0s

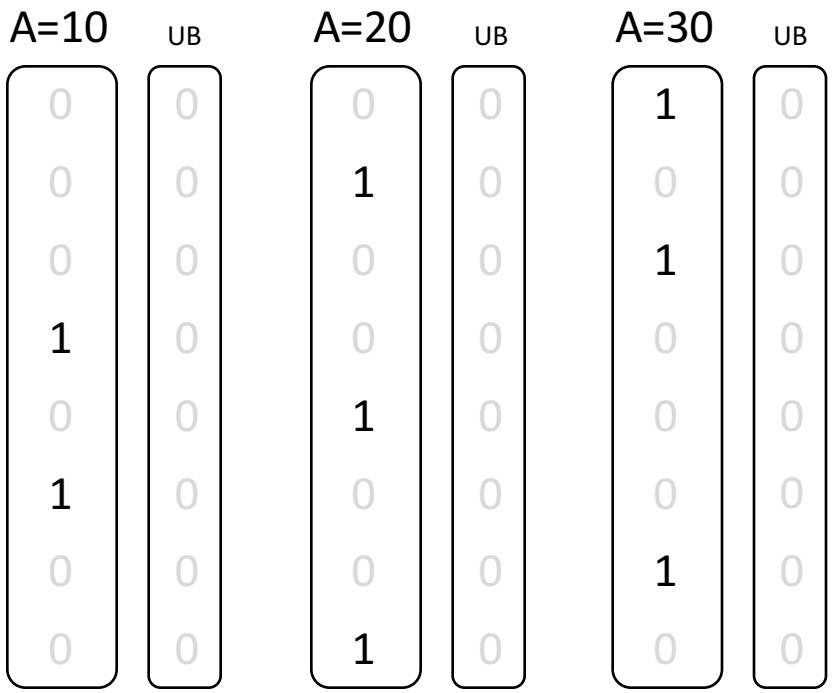
the current value is the XOR

every update flips a bit on UB

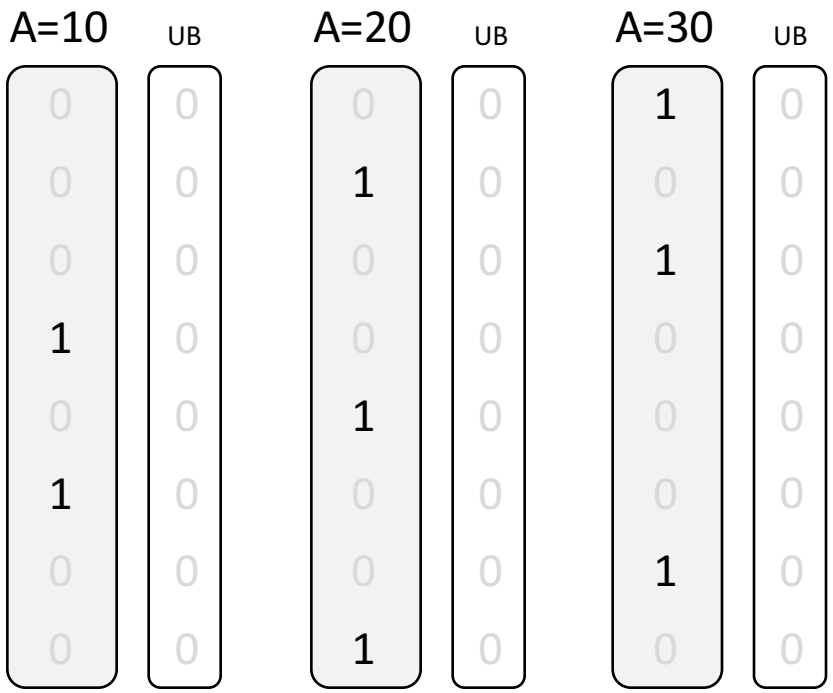
… distribute the update burden

Updating UpBit …

… row 2 to 10



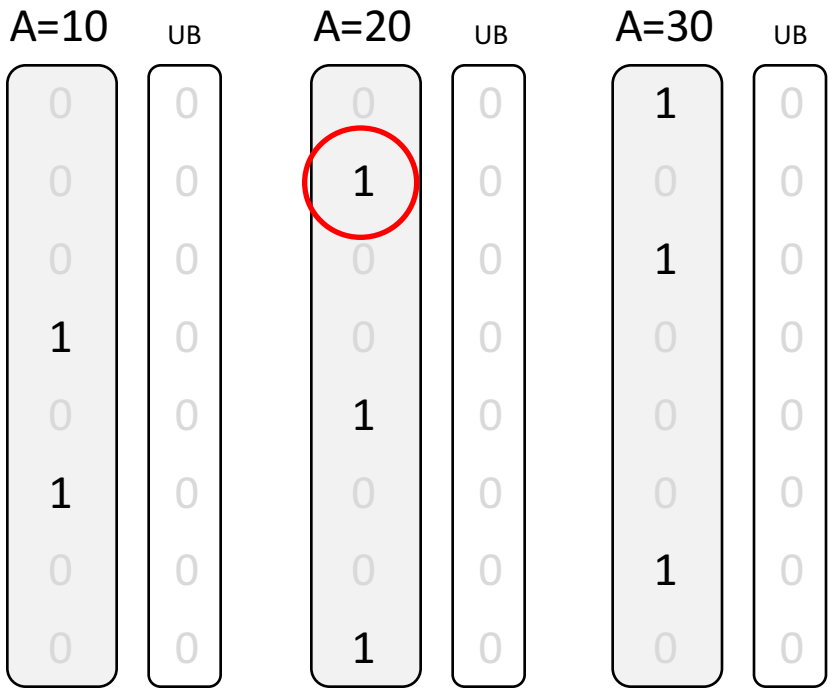
Updating UpBit …



… row 2 to 10

1. find old value of row 2 (A=20)

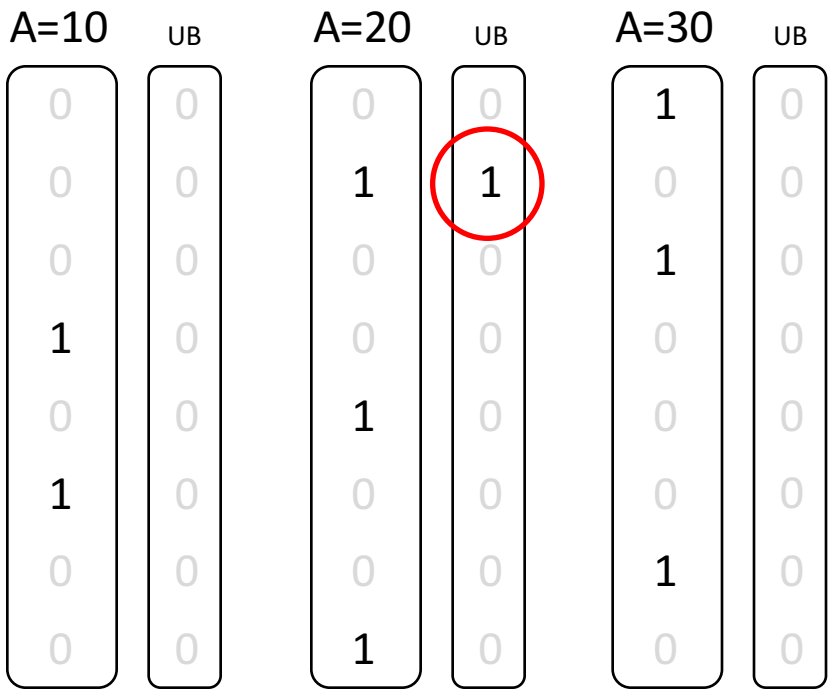
Updating UpBit …



… row 2 to 10

1. find old value of row 2 (A=20)

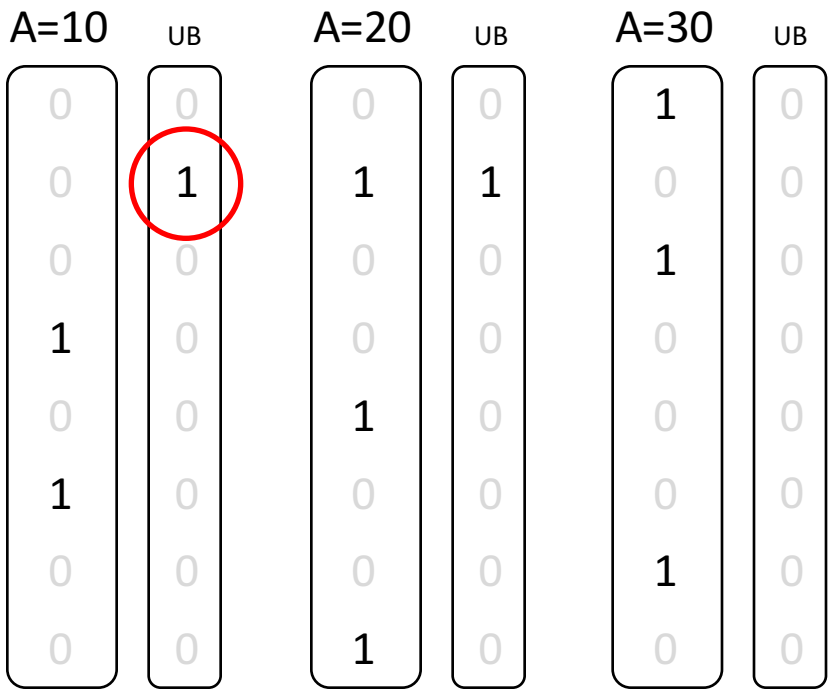
Updating UpBit …



… row 2 to 10

1. find old value of row 2 (A=20)
2. flip bit of row 2 of UB of A=20

Updating UpBit …



… row 2 to 10

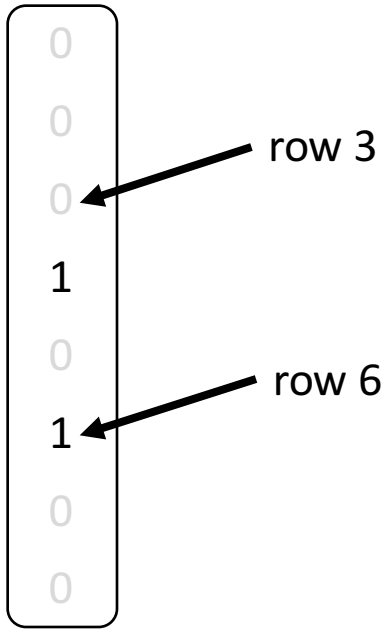
1. find old value of row 2 (A=20)
2. flip bit of row 2 of UB of A=20
3. flip bit of row 2 of UB of A=10

can we speed up step 1?

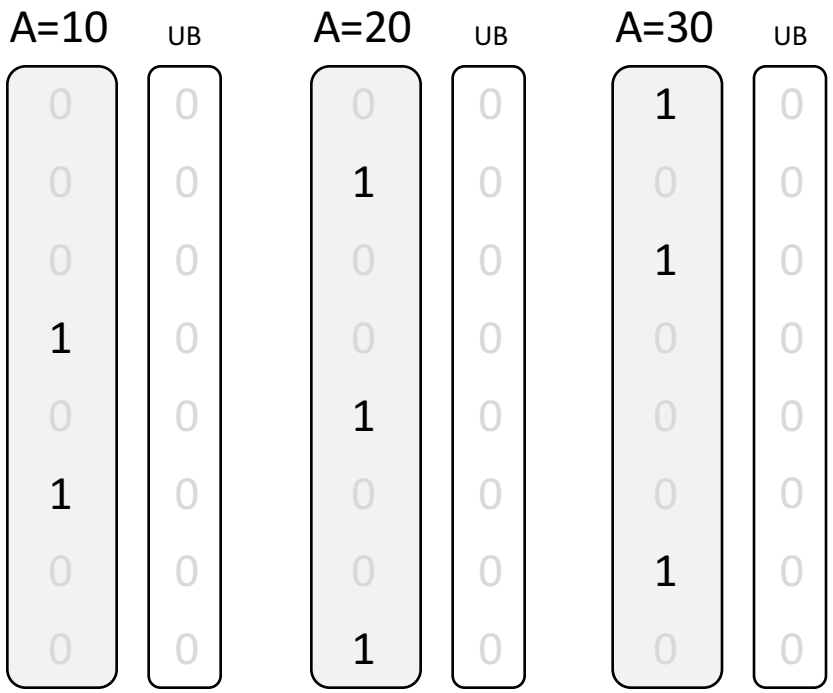
Design Element 2: fence pointers

efficient access of compressed bitvectors

fence pointers



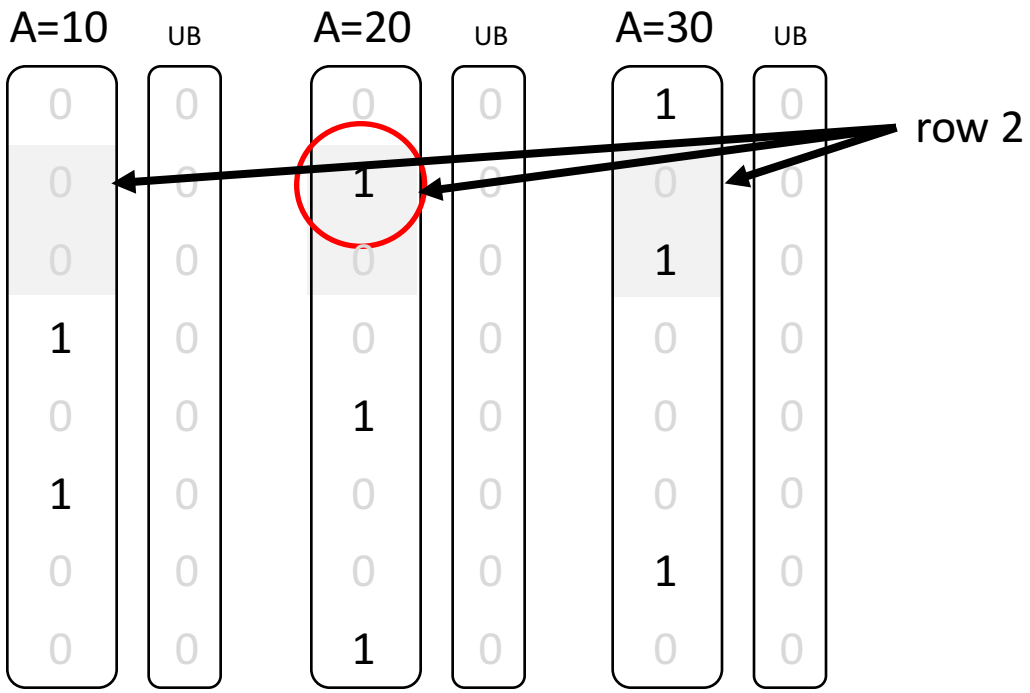
Updating UpBit …



… row 2 to 10

1. find old value of row 2 (A=20)

Updating UpBit (with fence pointers)…

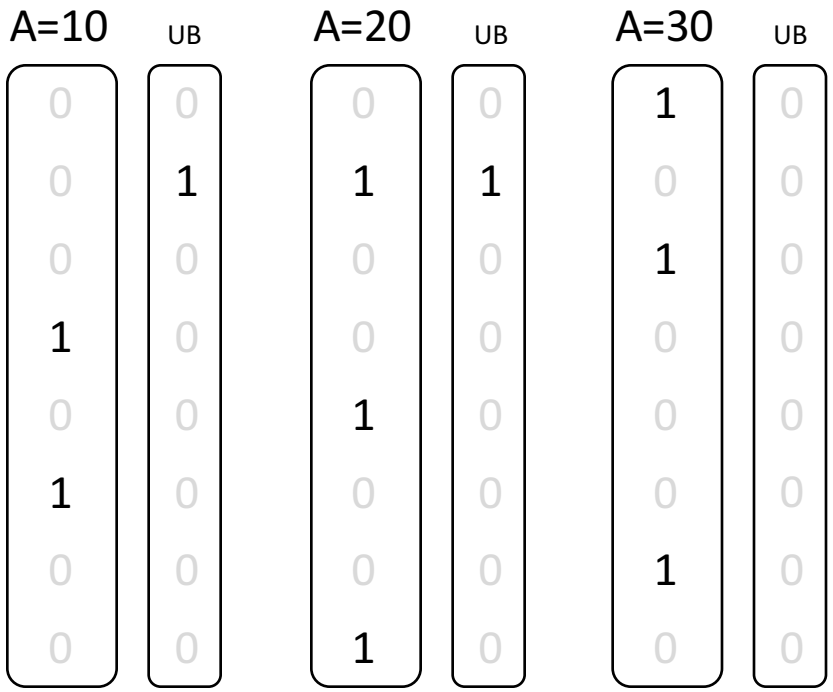


… row 2 to 10

1. find old value of row 2 (A=20)
   using fence pointers

Querying

Querying UpBit …



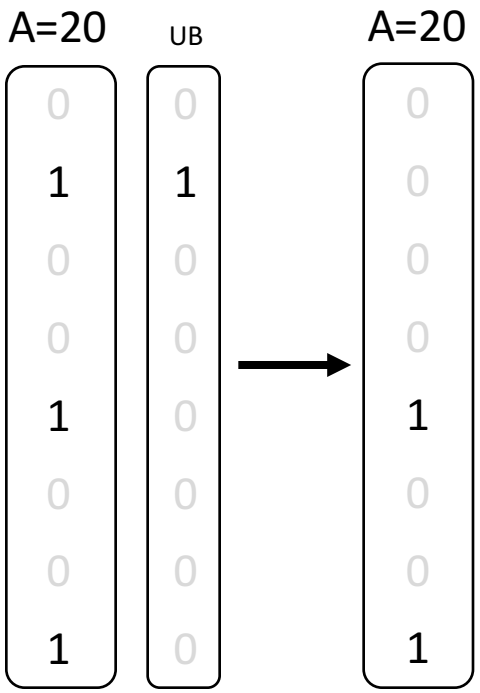
… A = 20

Return the XOR of A=20 and UB

Querying UpBit …

… A = 20

Return the XOR of A=20 and UB

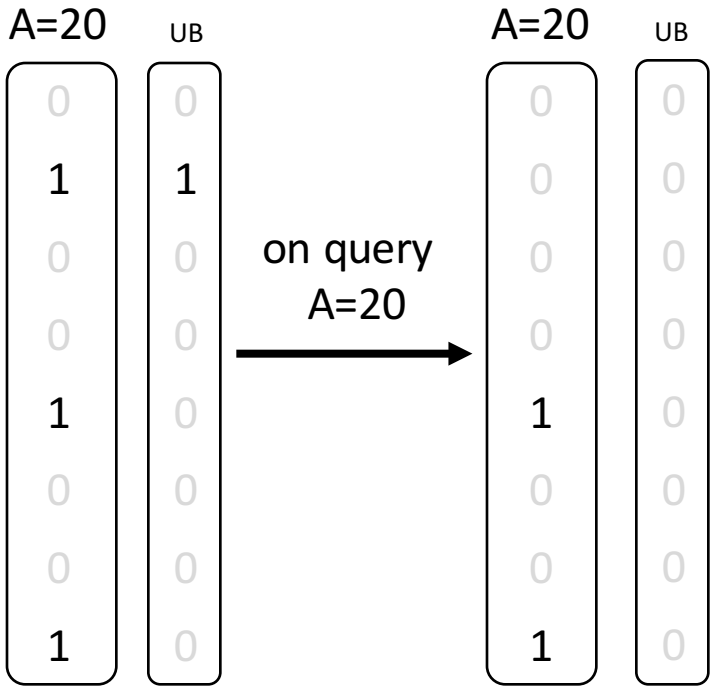


can we re-use the result?

Design Element 3: query-driven merging

maintain high compressibility of UB

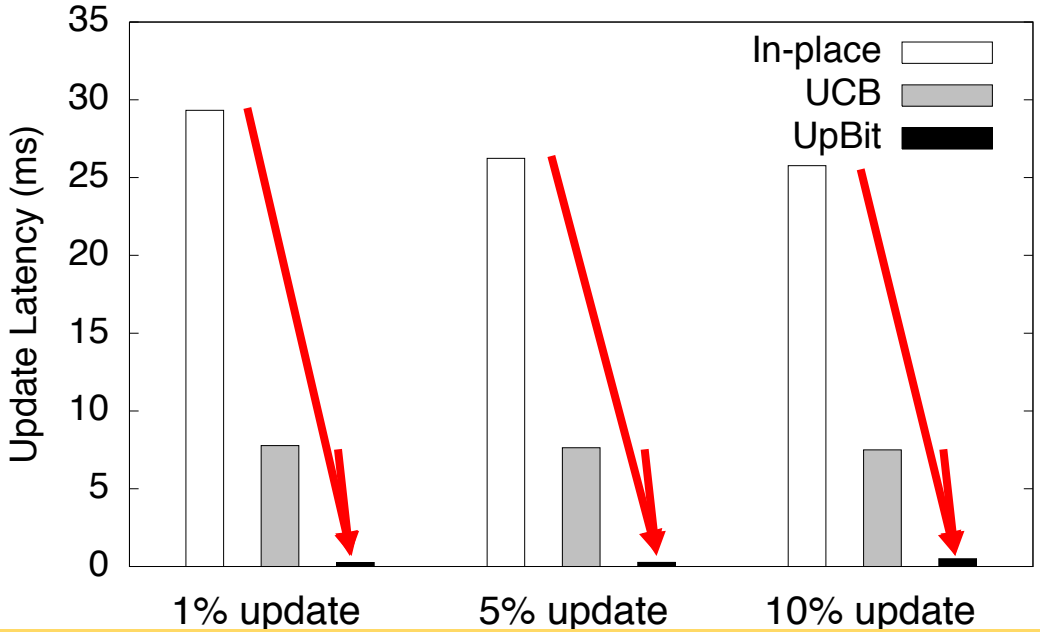
query-driven merging

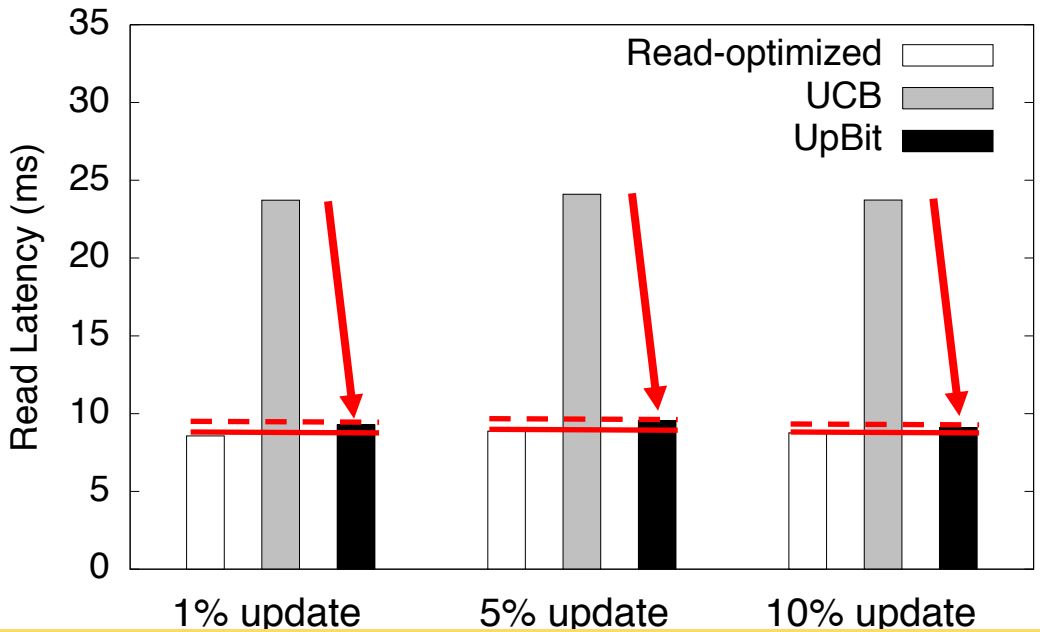


UpBit supports very efficient updates

n=100M tuples, d=100 domain values

100k queries (varying % of updates)





updates: 15-29x faster than UCB

only 8% read overhead over optimal

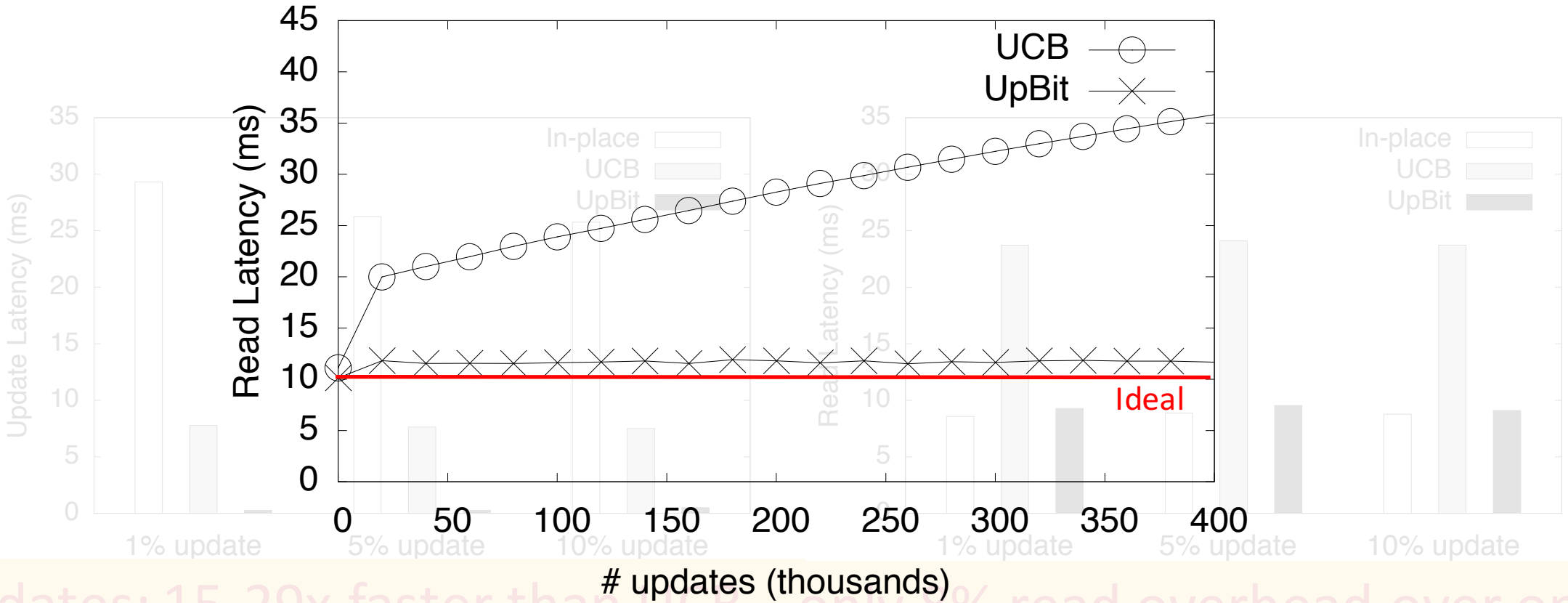
51-115x faster than in-place

3x faster reads than UCB

UpBit offers robust reads

n=100M tuples, d=100 domain values

50%/50% update/read queries



updates: 15-29x faster than UCB

only 8% read overhead over optimal

51-115x faster than in-place

3x faster reads than UCB

More in the paper …

Tuning: how frequent to merge UB to the index?

Tuning: what is the optimal granularity of fence pointers?

Optimizations: multi-threaded reads and updates

Performance: full query analysis (scientific data and TPCH)

UpBit: achieving scalable updates

distribute the update burden

update bitvectors

efficient bitvector accesses

fence pointers

avoid redundant bitwise operations

query-driven merging of UB

Thanks!

DASlab

FNSNF

http://daslab.seas.harvard.edu/rum/