

Dynamic Searchable Encryption with Small Client Storage

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What is Dynamic Searchable Encryption (DSE)?

Client



\mathcal{L}^{Query} leakage---Search pattern: whether a search query is repeated

\mathcal{L}^{Setup} leakage: total leakage prior to query execution e.g. size of each encrypted file, size of the encrypted index

Untrusted Cloud

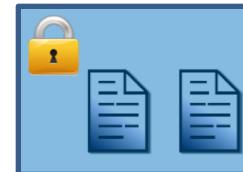


search query:

keyword

\mathcal{L}^{Query} leakage---Access

pattern: encrypted document ids and files that satisfy the search query



\mathcal{L}^{Update} leakage:

leakage during update execution

update query:

keyword

Security (informal): The adversary does not learn anything beyond the above leakages!

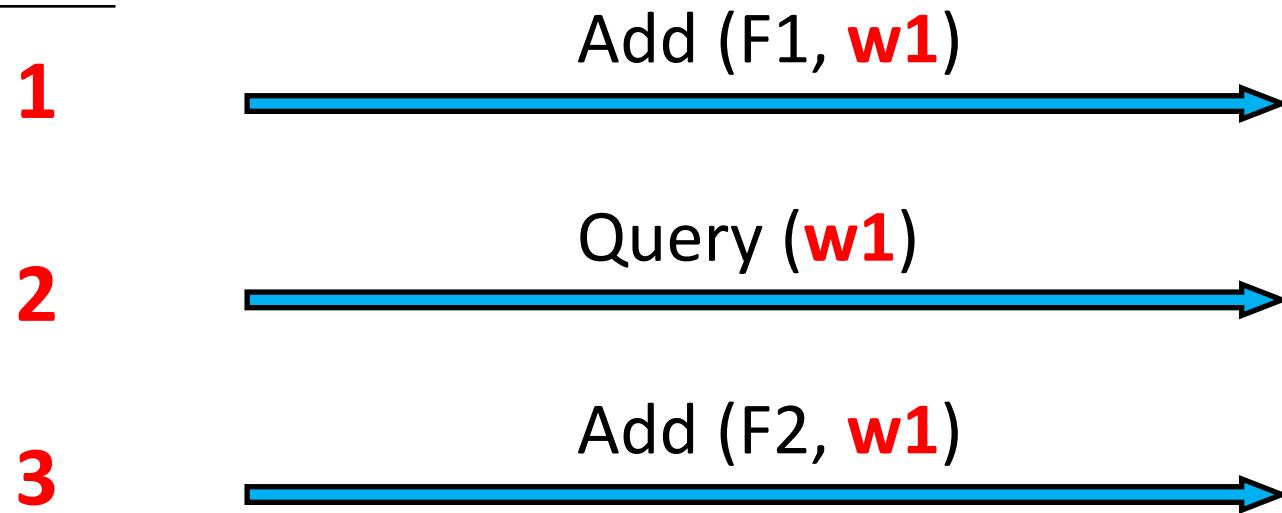
Update Leakage --- Forward Privacy



High-level idea: The server should not be able to relate an update with a previous operation!



Time



- Server should **not** learn that update in timestamp **3** is for the same keyword!
- **Definition:** A DSE scheme is forward private if the update does not reveal any information about the involved keyword, i.e., $\mathcal{L}^{Update}(w) = \mathcal{L}(op, id)$

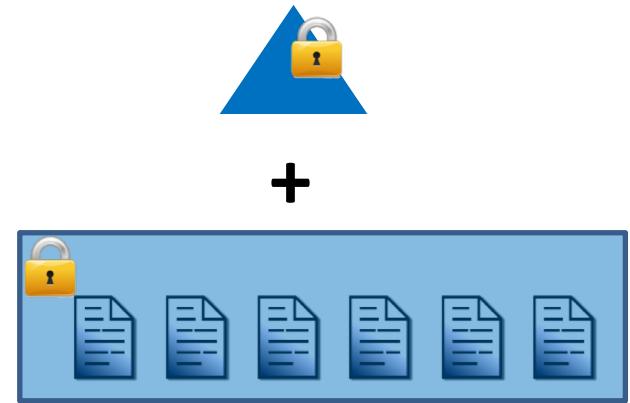
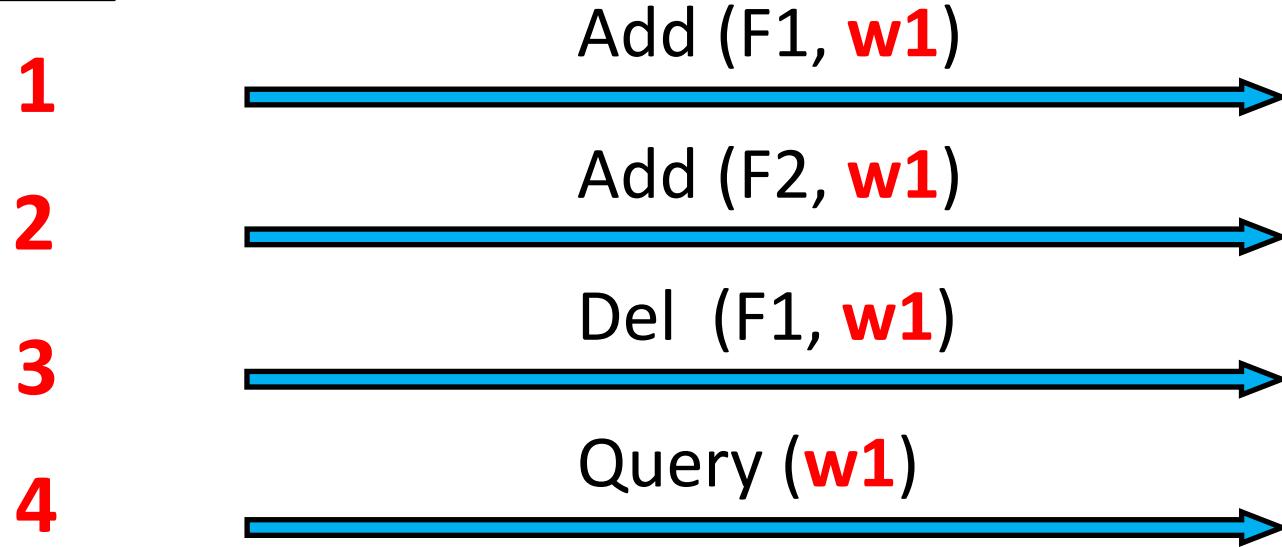
Update Leakage --- Backward Privacy

Client



High-level idea: The server should reveal controlled information about deleted files during search

Time



$$TimeDB(w_1) = \{(F2, 2)\}$$

$TimeDB(w)$ = {files **currently** containing w and when they were stored}

$Updates(w)$ = {timestamp of each update for w } $Updates(w_1) = \{1, 2, 3\}$

$DelHist(w)$ = {exactly which deletion cancels which addition} $DelHist(w_1) = \{(1, 3)\}$

Update Leakage --- Backward Privacy

Client



High-level idea: The server should reveal controlled information about deleted files during search

Time

1

Add (F1, **w1**)

2

Add (F2, **w1**)

3

Del (F1, **w1**)

4

Query (**w1**)



+



Backward Privacy Type – I: $\mathcal{L}^{Search}(w) = \mathcal{L}(\text{TimeDB}(w))$

Backward Privacy Type – II: $\mathcal{L}^{Search}(w) = \mathcal{L}(\text{TimeDB}(w), \text{Updates}(w))$

Backward Privacy Type – III: $\mathcal{L}^{Search}(w) = \mathcal{L}(\text{TimeDB}(w), \text{Updates}(w), \text{DelHist}(w))$

More Leakage ↓

Issues with Prior Forward & Backward DSE schemes



Require: The client to store an operation counter for each unique keyword!!!



Keyword	Counter
w1	3
w2	2
w3	4
...	
wM	5

$O(|W|)$ can be up to $O(N)$, where N is the total DB size

$O(|W|)$, where $|W|$ is the dictionary size

Synchronization issues: if the client wants to access the encrypted DB from multiple devices



Issues with Prior Forward & Backward DSE schemes

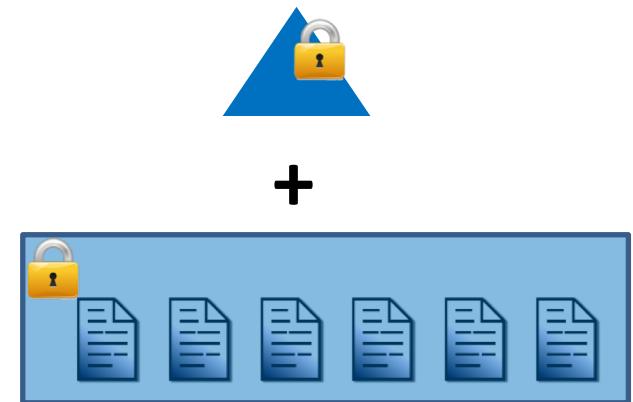


Client **Outsourcing** the operational counters to the server requires the use of Oblivious Indexes ~**extra $O(\log^2 N)$** overhead and **$O(\log N)$** rounds of interaction!



E.g., an Oblivious Hash Map (**OMAP**)

Keyword	Counter
w1	3
w2	2
w3	4
...	
wM	5



Prior state-of-the-art Works & Our Contributions

	Search	Update	Search RT	BP-Type
Moneta	$\tilde{O}(a_w \log N + \log^3 N)$	$\tilde{O}(\log^2 N)$	2	Type-I
OMAP+Mitra	$O(a_w + \log^2 N)$	$O(\log^2 N)$	$O(\log N)$	Type-II
SDa	$O(a_w + \log N)$	$O(\log N)$ (am.)	1	Type-II
SDd	$O(a_w + \log N)$	$O(\log^3 N)$	1	Type-II
ORION	$O(n_w \log^2 N)$	$O(\log^2 N)$	$O(\log N)$	Type-I
HORUS	$O(n_w \log d_w \log N + \log^2 N)$	$O(\log^2 N)$	$O(\log N)$	Type-III
QOS	$O(n_w \log i_w + \log^2 N)$	$O(\log^3 N)$	$O(\log N)$	Type-III

N : total number of (file, keyword) pairs, a_w : #updates for keyword w

n_w : #files **currently** containing keyword w , i_w : #inserts for keyword w

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SDa and **SDd** have **20x-85x** faster search time than MITRA
QOS has **14x-16531x** faster search time than HORUS

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QOS	$O(n_w \log i_w + \log^2 N)$	$O(\log^3 N)$	$O(\log N)$	Type-III

QOS has better search time for deletion ratios between **40%-80%**

SDa scheme

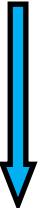
Client



High-level idea: Organize N updates in a collection of at most $\log_2 N$ independent encrypted indexes



Add (F_1, w_1)



SDa scheme

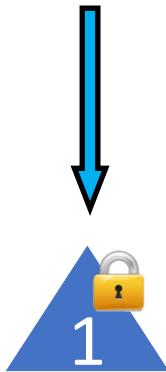
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High-level idea: Organize N updates in a collection of at most $\log_2 N$ independent encrypted indexes



Add (F4, w1)



SDa scheme

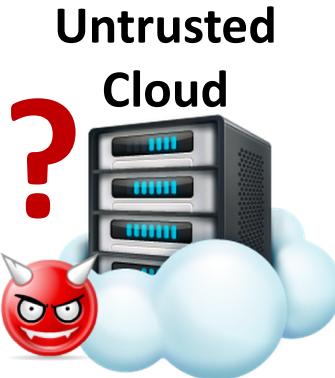


Add (F4, w1)



High-level idea: Organize N updates in a collection of at most $\log_2 N$ independent encrypted indexes

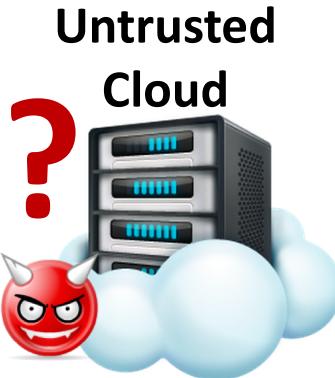
If we keep adding the number of encrypted indexes will be $O(N)$



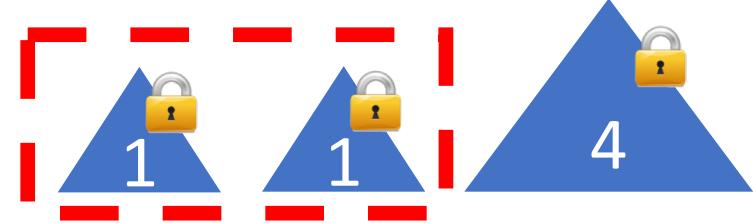
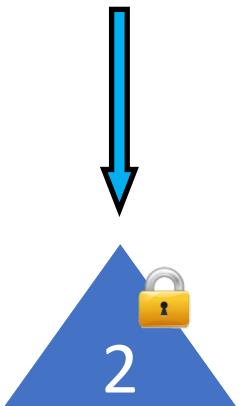
SDa scheme



High-level idea: Organize N updates in a collection of at most $\log_2 N$ independent encrypted indexes



Whenever **two** indexes of the same size exist, **download them and merge them in a new index**



SDa scheme

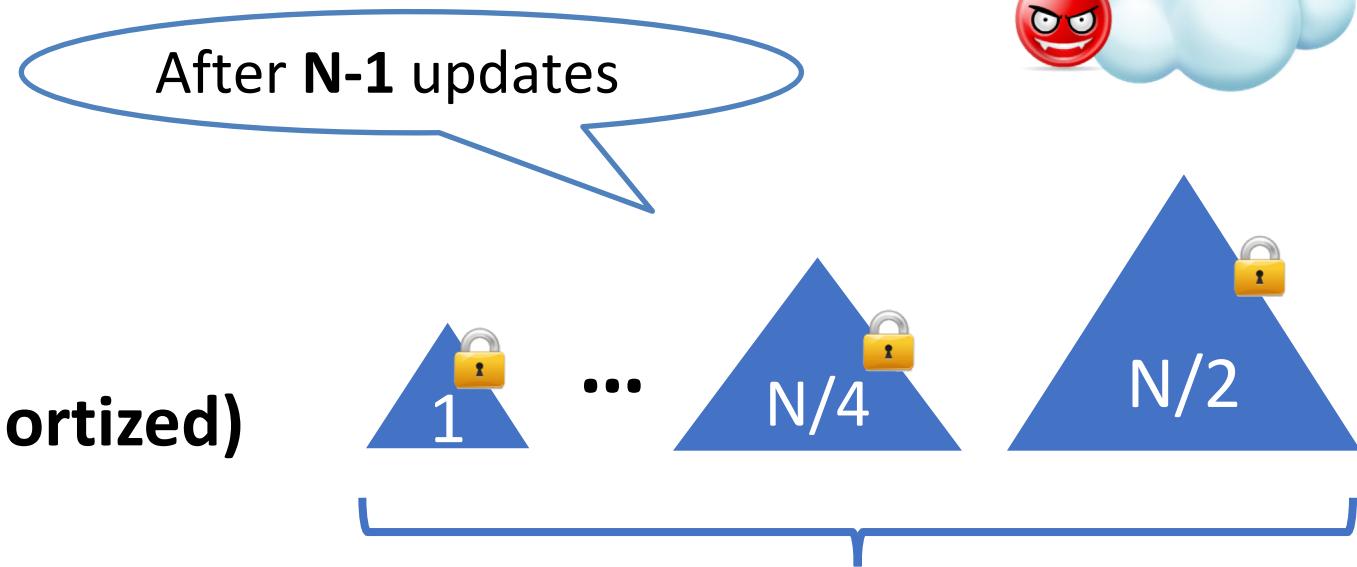


High-level idea: Organize N updates in a collection of at most $\log_2 N$ independent encrypted indexes

*Assuming that N is a power of 2



Update cost = $O(\log_2 N)$ (amortized)



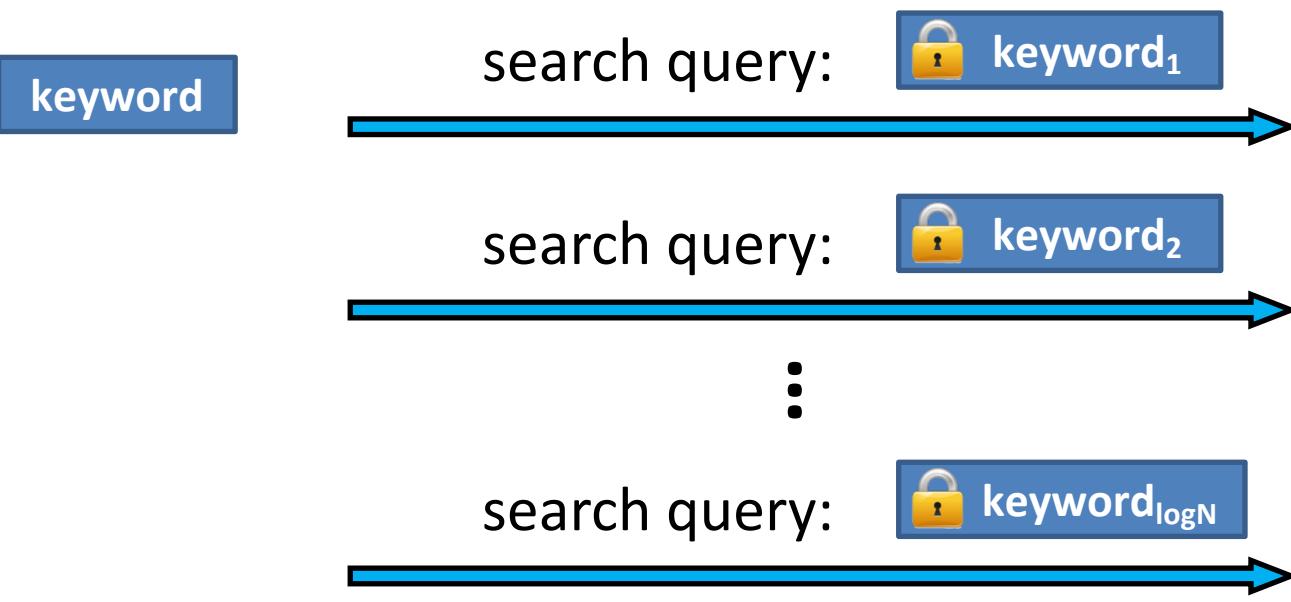
$O(\log_2 N)$
encrypted indexes

SDa scheme

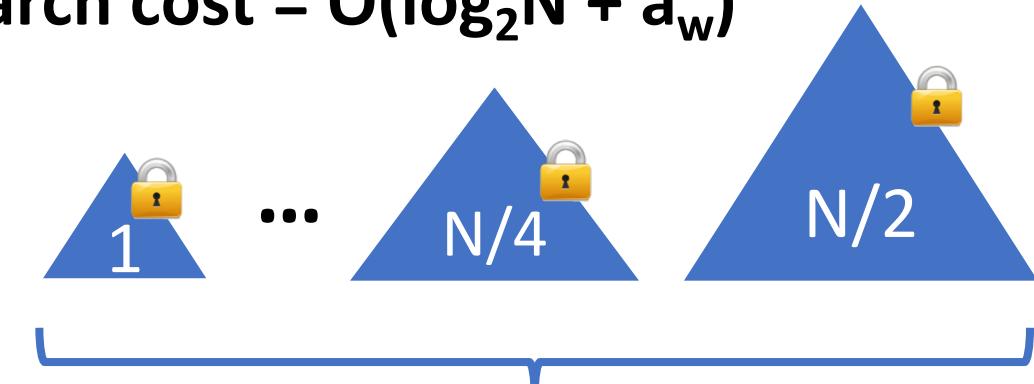


High-level idea: Organize N updates in a collection of at most $\log_2 N$ independent encrypted indexes

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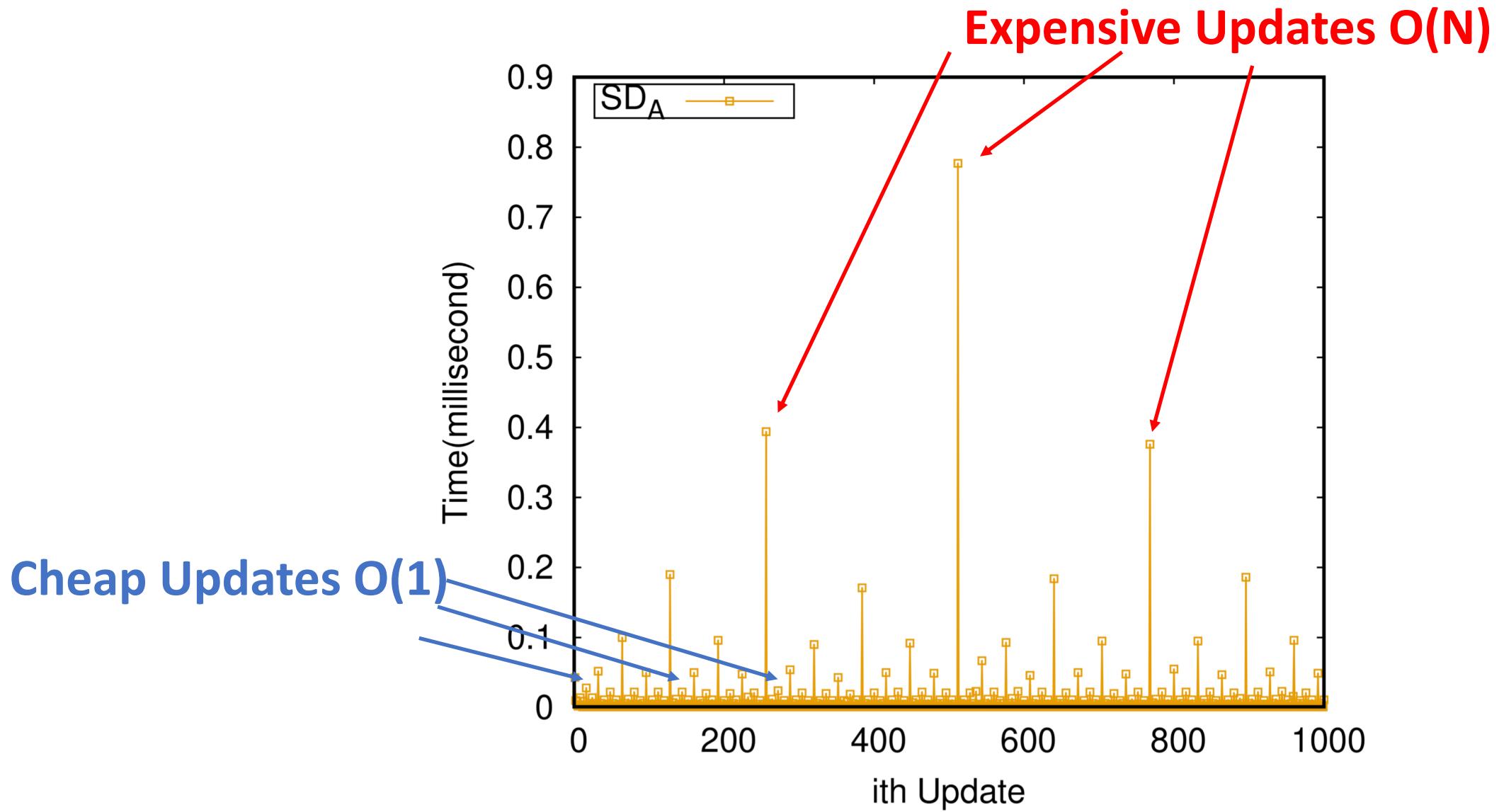


$$\text{Search cost} = O(\log_2 N + a_w)$$



Intuition Forward/Backward privacy: Every index is built with a fresh key and the used static SE is response hiding!!!

SDa --- Amortized Update Cost



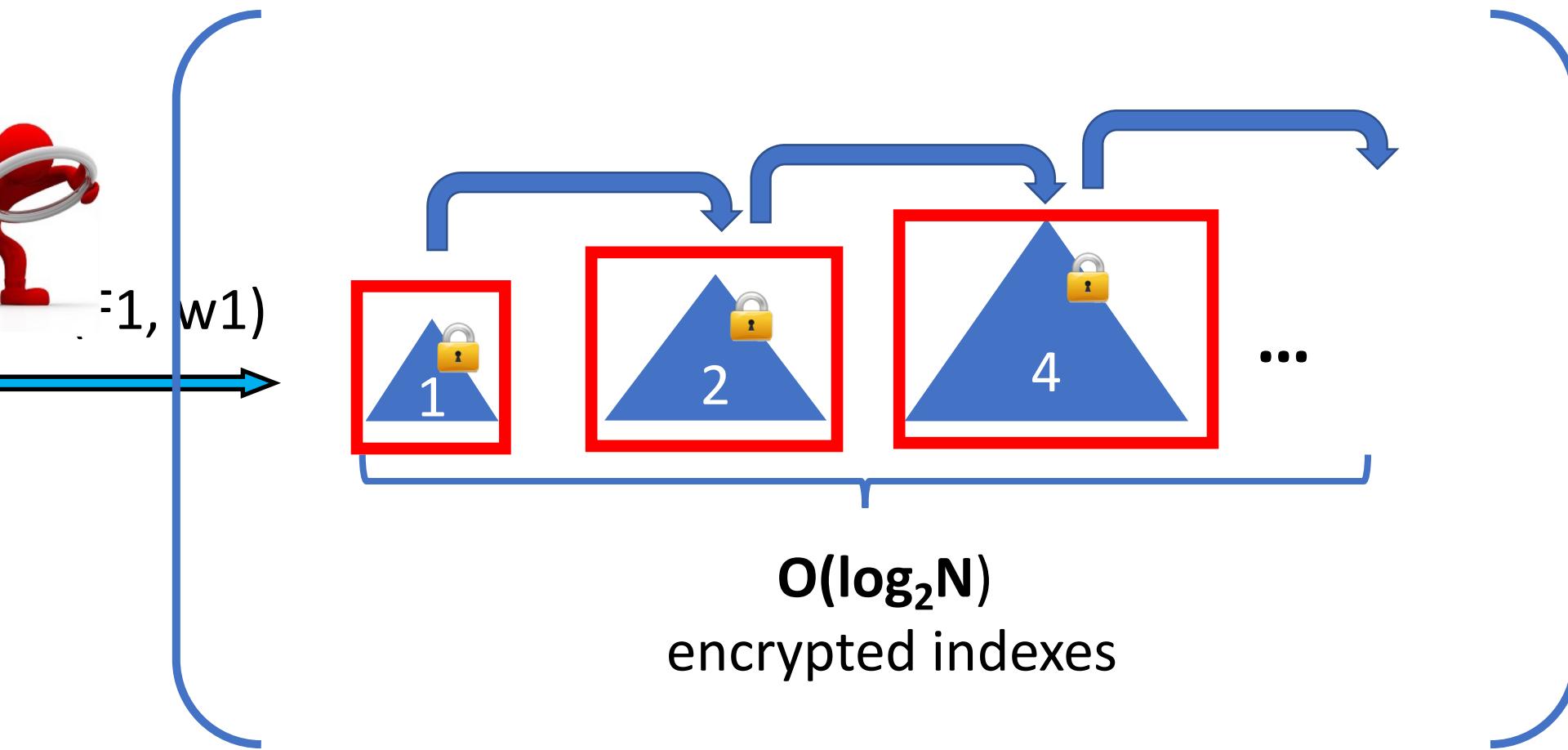
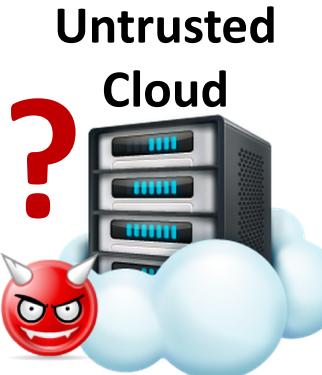
SDd scheme

Client



$(-1, w_1)$

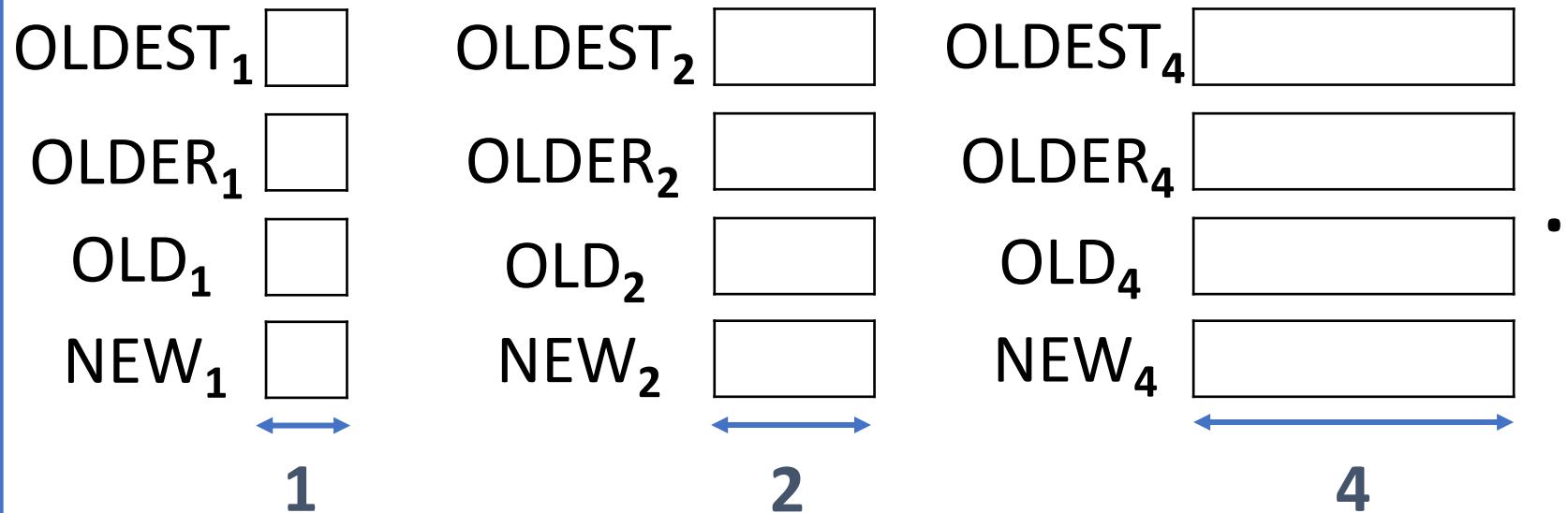
High-level idea: Let's **de-amortize** the SDa construction



$O(\log_2 N)$
encrypted indexes

SDd scheme

Client

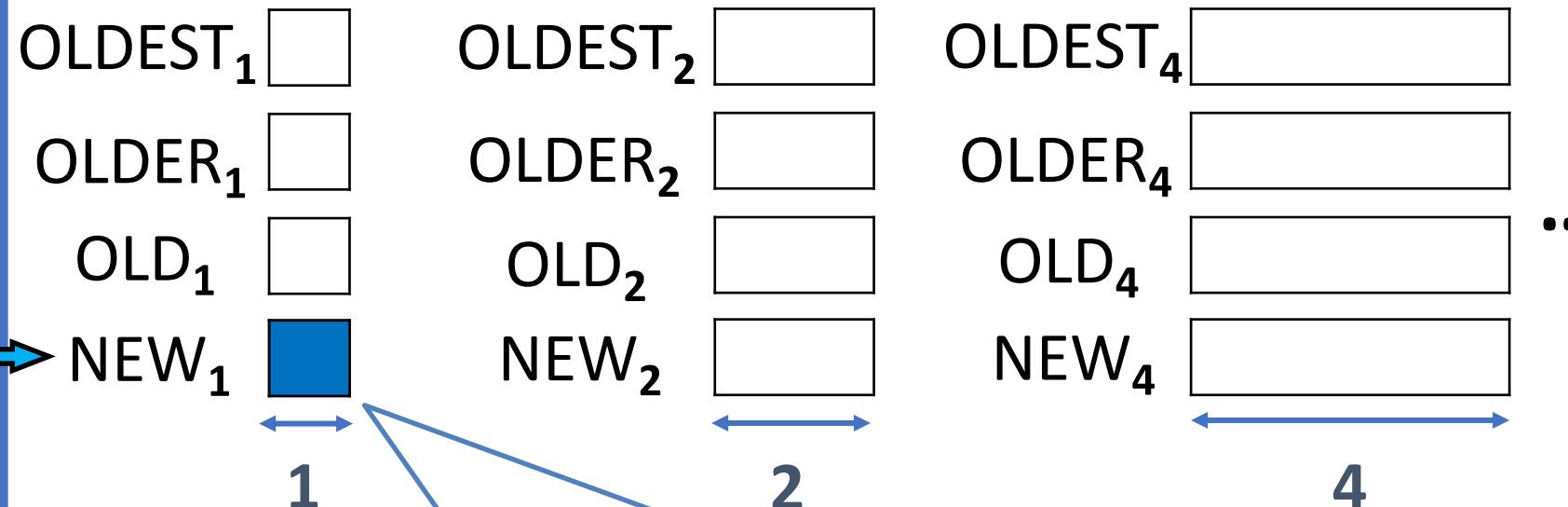


SDd scheme

Client



Add (F_1, w_1)



If **NEW** is full move it to the first
empty OLDEST, OLDER or OLD index

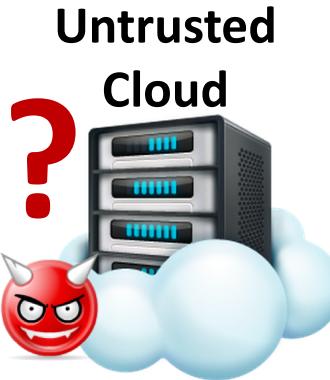
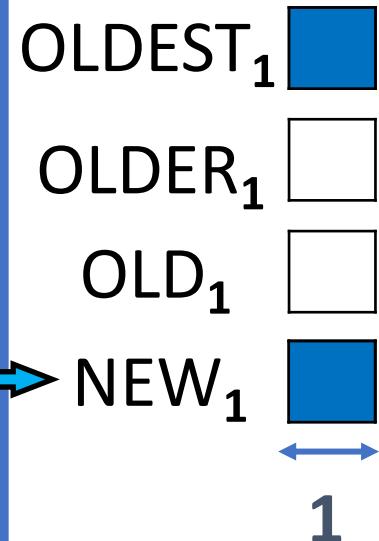


SDd scheme

Client



Add (F_2, w_1)

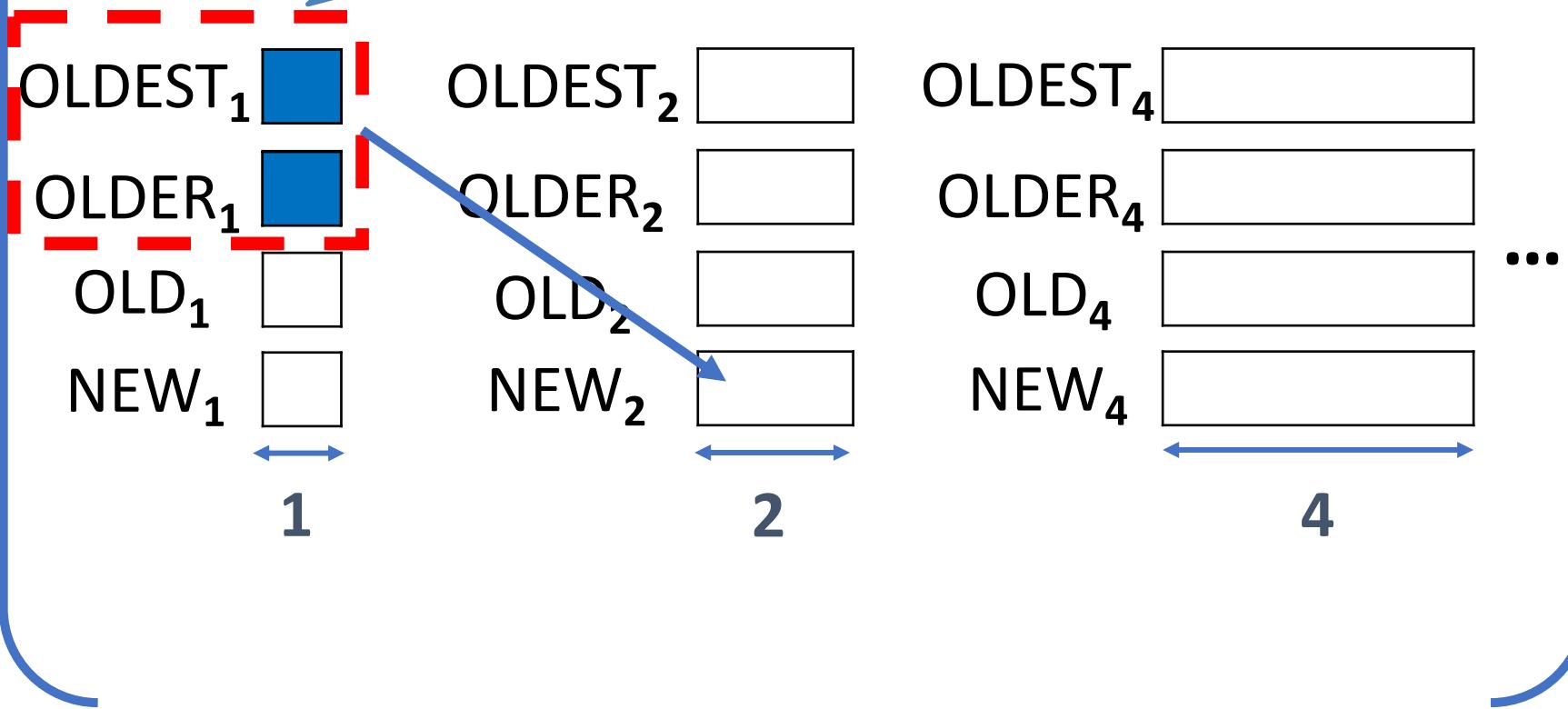


SDd scheme

Client



If OLDEST_i and OLDER_i are full start moving the updates to the NEW_{i+1} index

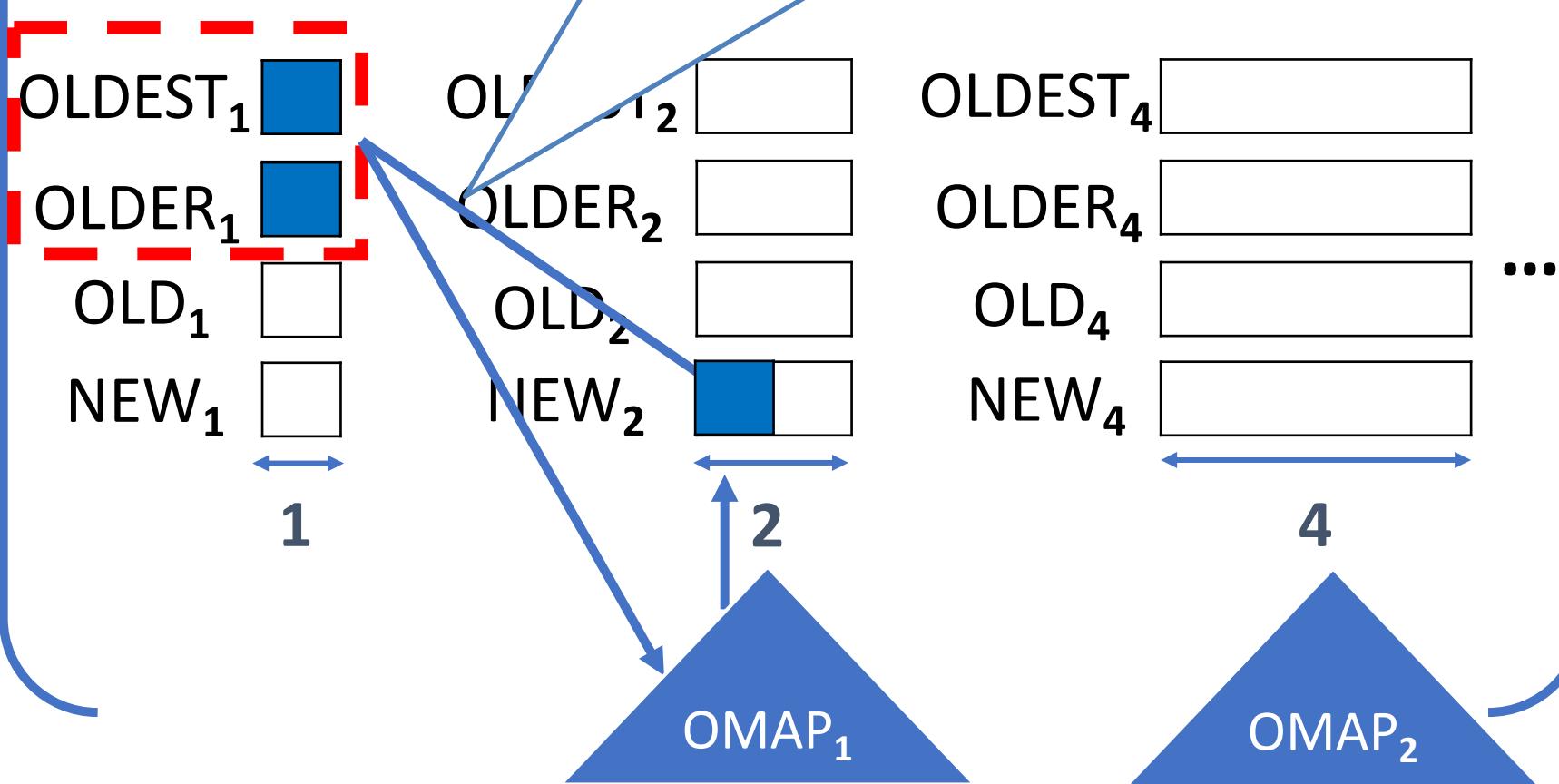


SDd scheme

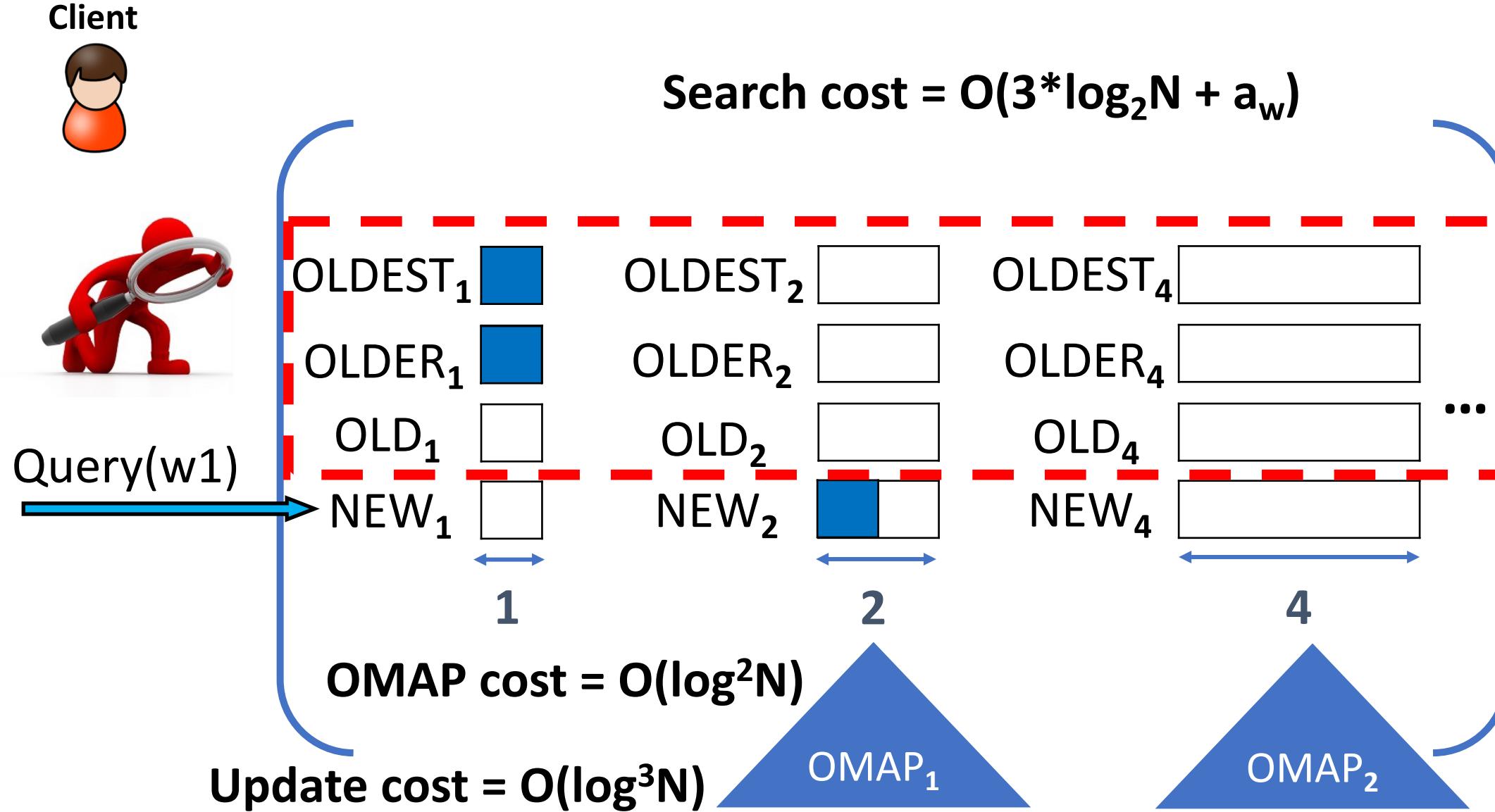
Client



For achieving **Forward Privacy** we need
to use Oblivious MAPs (OMAP)



SDd scheme



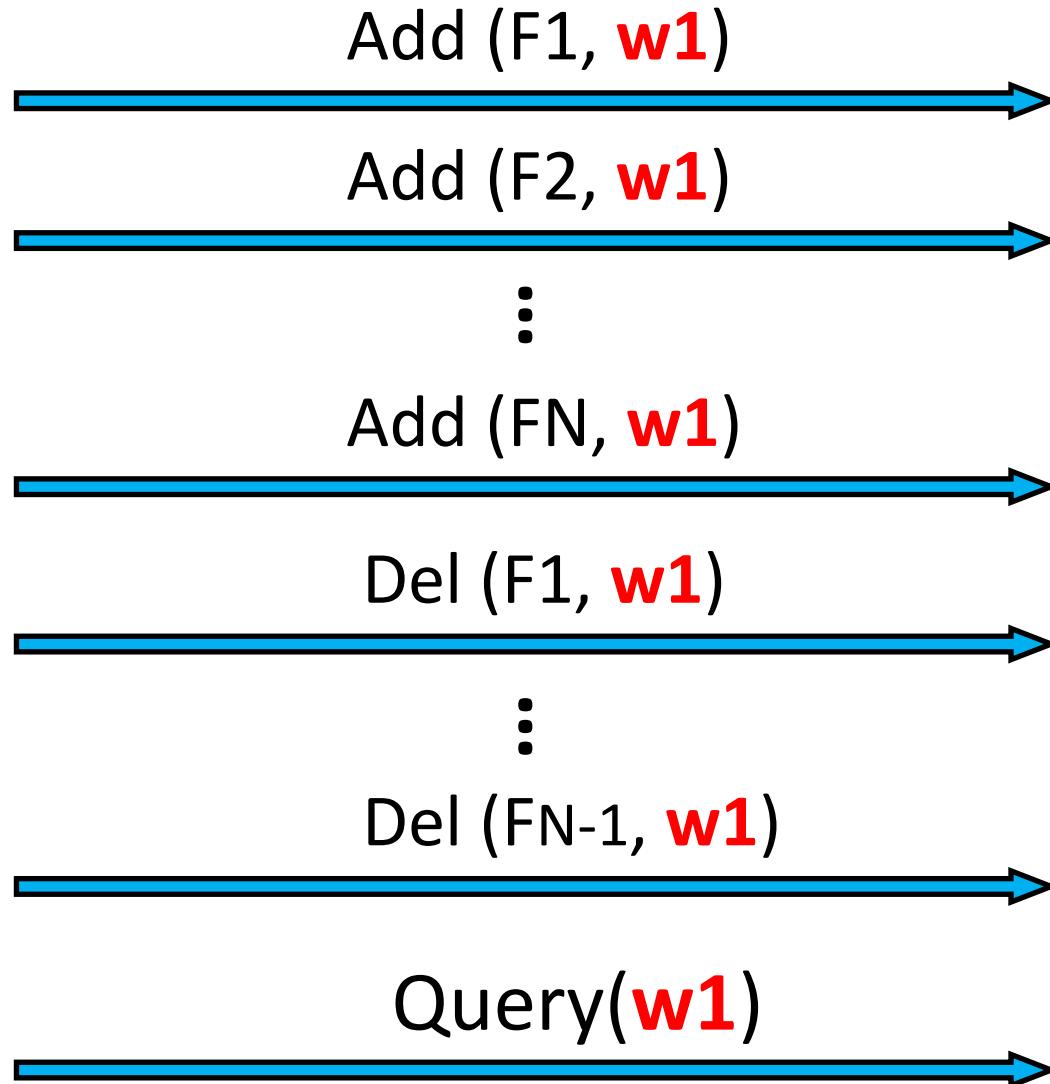
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QOS	$O(n_w \log i_w + \log^2 N)$	$O(\log^3 N)$	$O(\log N)$	Type-III

- **Definition:** A DSE scheme has optimal (resp. quasi-optimal) search time, if the asymptotic complexity of Search is $O(n_w)$ (resp. $O(n_w \text{polylog}(N))$).

Motivation for Optimal/Quasi-optimal Search

Client



SDd search cost
= $O(\log_2 N + a_w)$



+



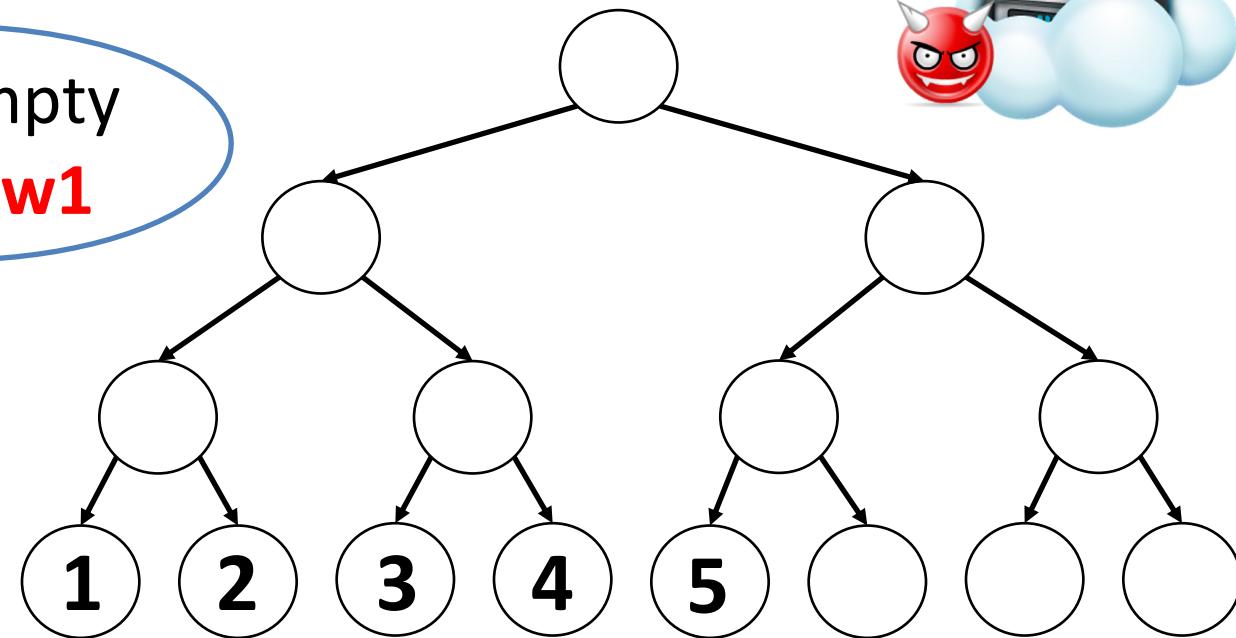
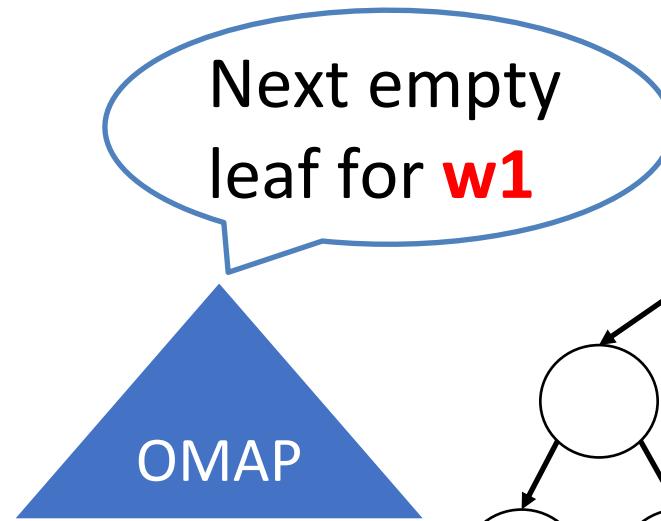
w1 is contained only in
File N, but the result has
size $O(a_w) \approx O(N)$

QOS --- Main Idea



Idea: For each keyword w create a data structure that helps us avoid the deleted regions

- Add (F1, $w1$)
- Add (F2, $w1$)
- Add (F5, $w1$)
- Add (F10, $w1$)
- Add (F35, $w1$)



Tree for keyword $w1$ ($N=8$)



QOS --- Main Idea

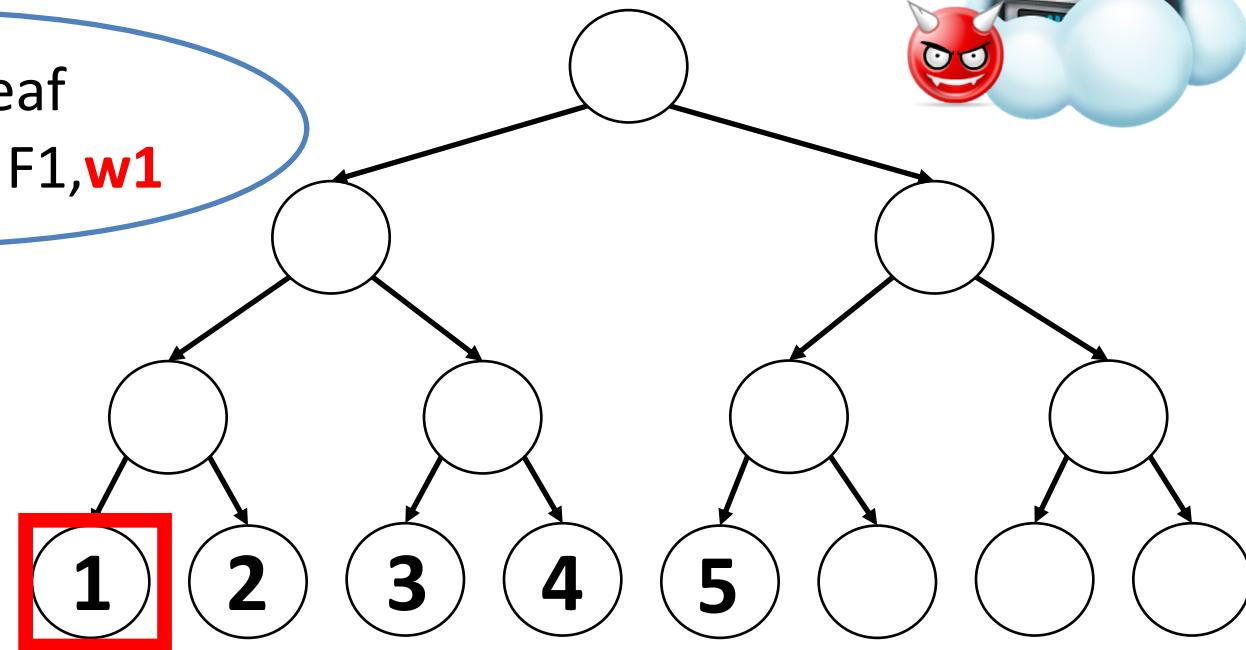
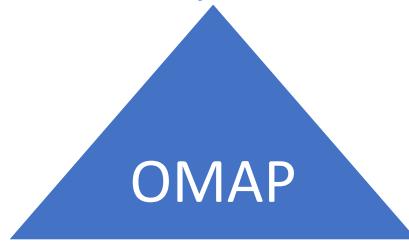


Idea: For each keyword w create a data structure that helps us avoid the deleted regions



Del (F1, w_1)

Returns the leaf
that contains F1, w_1



Tree for keyword w_1 ($N=8$)

QOS --- Main Idea

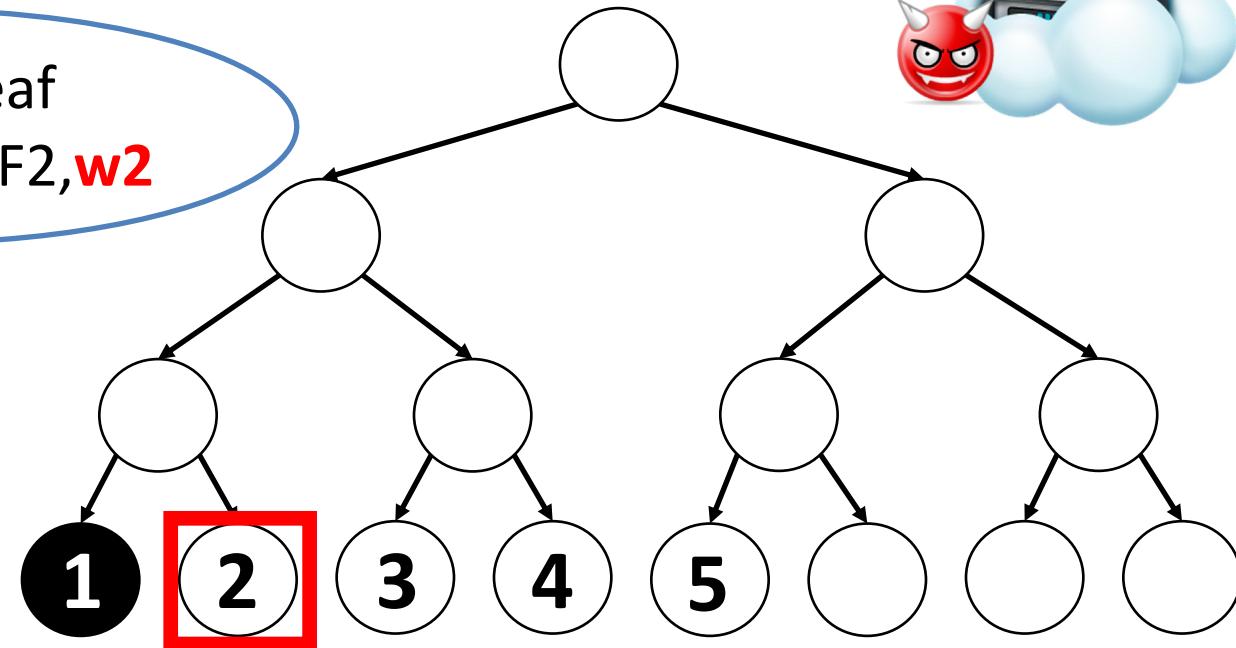
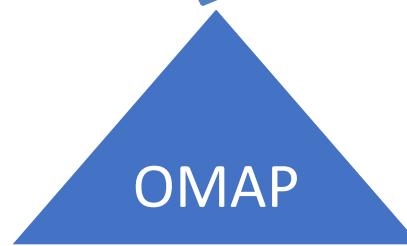


Idea: For each keyword w create a data structure that helps us avoid the deleted regions



Del (F1, w_1)
Del (F2, w_1)

Returns the leaf
that contains F2, w_2

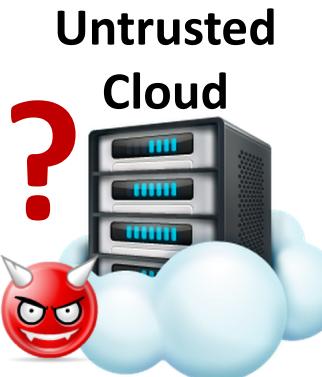


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QOS --- Main Idea

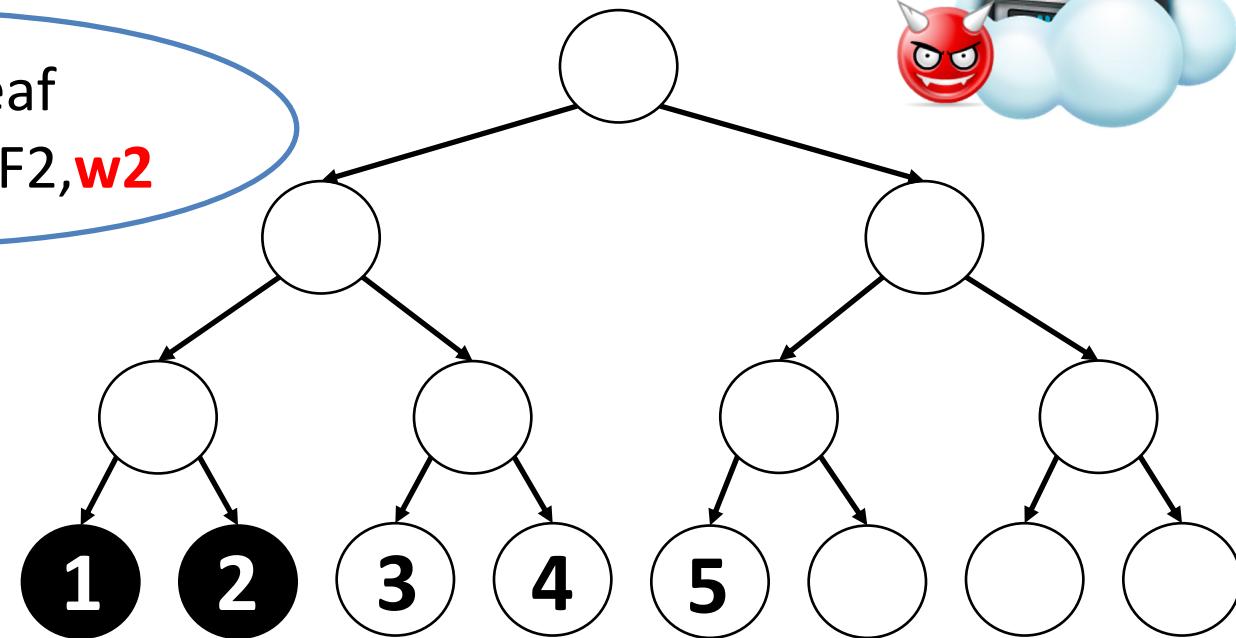
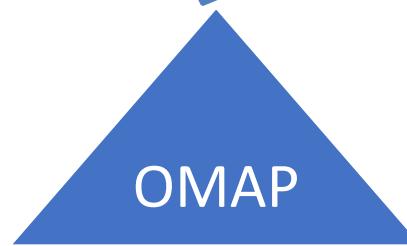


Idea: For each keyword w create a data structure that helps us avoid the deleted regions



Del (F1, w_1)
Del (F2, w_1)

Returns the leaf
that contains F2, w_2



Tree for keyword w_1 ($N=8$)

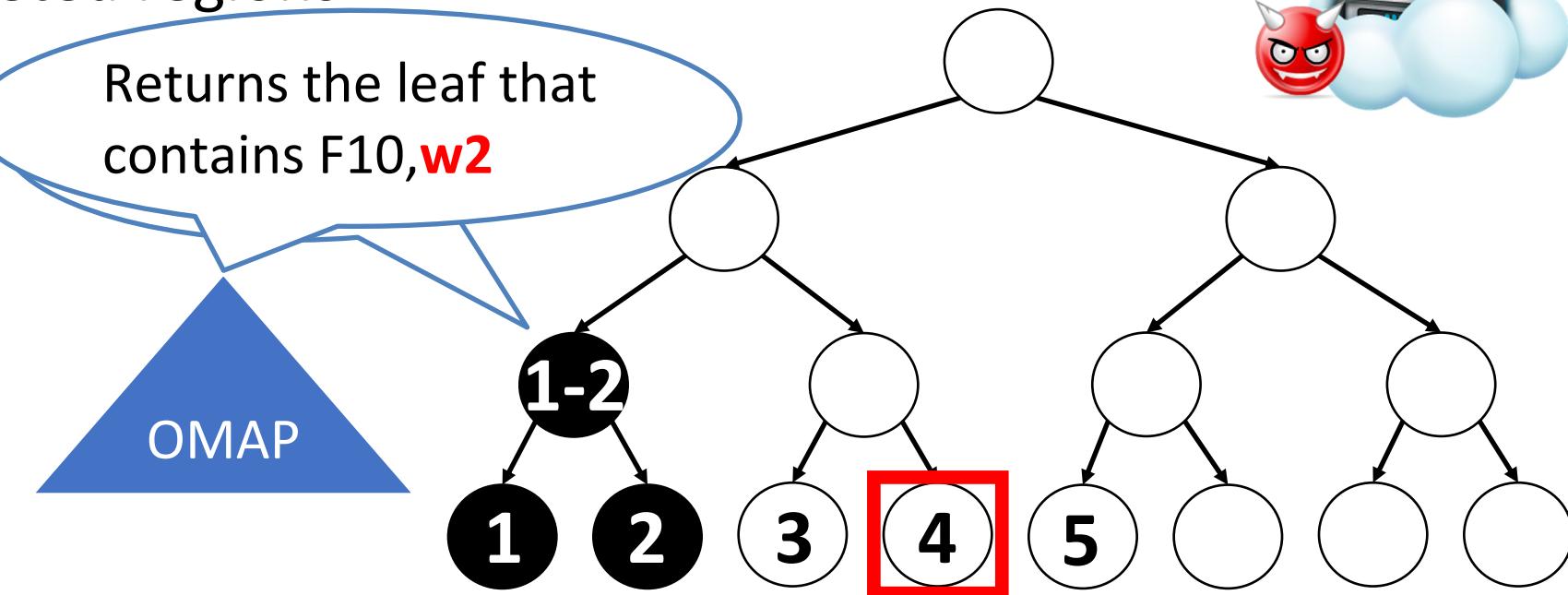
QOS --- Main Idea



Idea: For each keyword w create a data structure that helps us avoid the deleted regions



Del (F1, w_1)
Del (F2, w_1)
Del (F10, w_1)



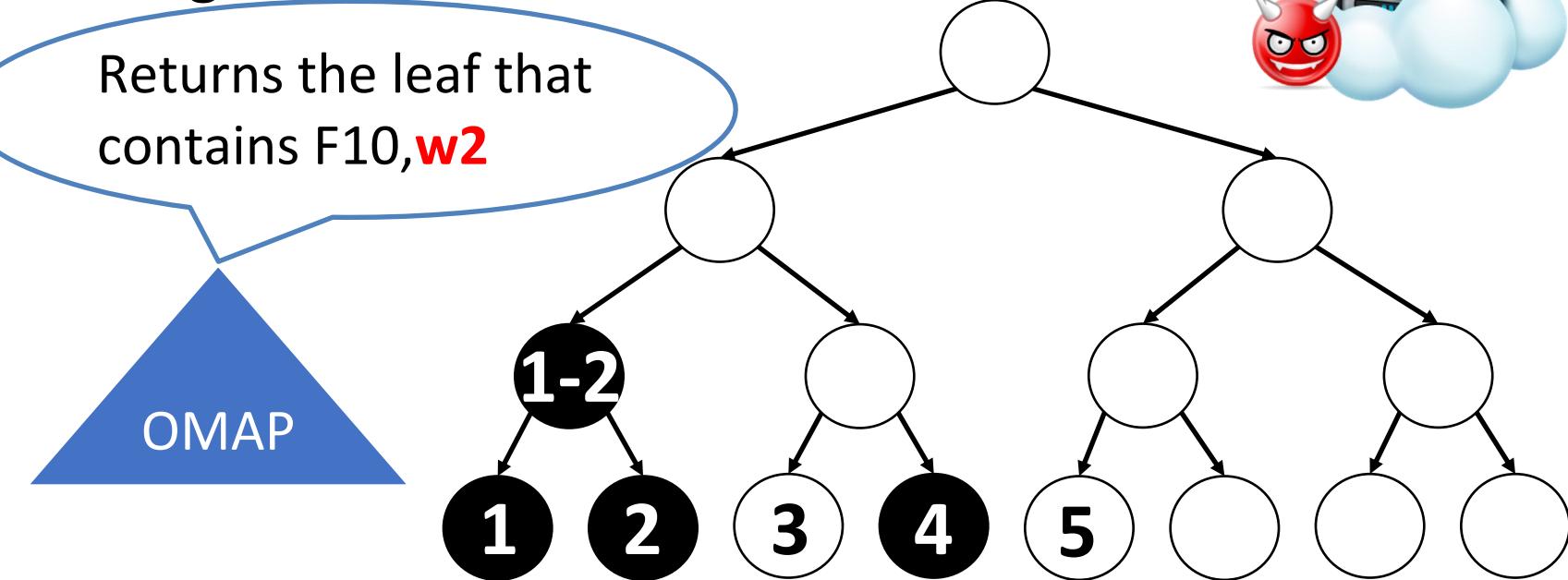
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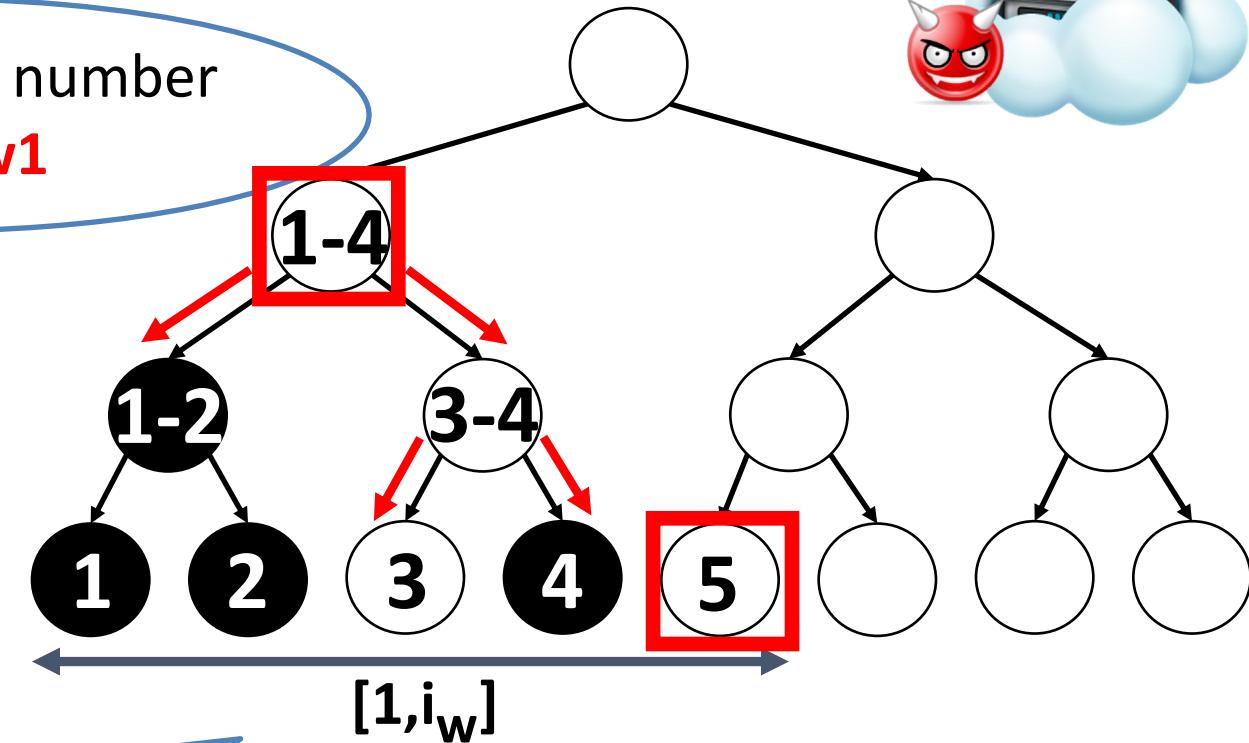


Query(w_1)

Returns i_w the number of inserts for w_1

OMAP

Search cost = $O(\log^2 N + n_w \log i_w)$



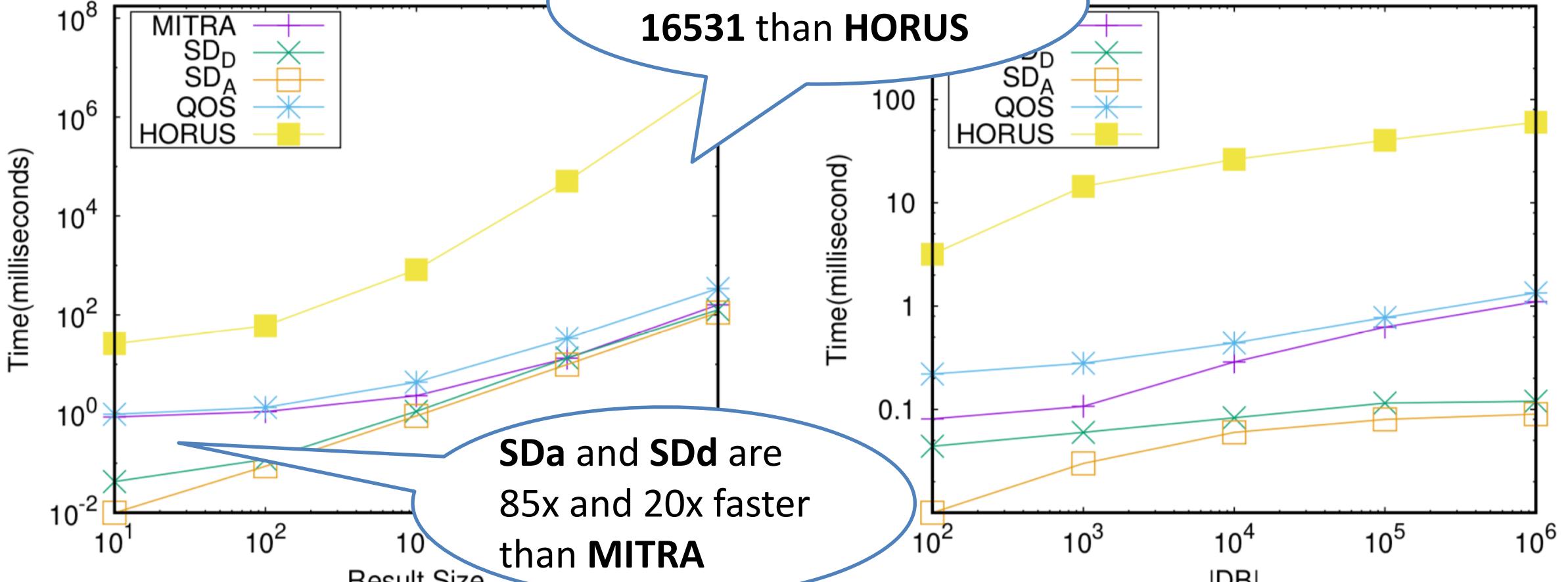
Computes the Best
Range Cover of $[1, i_w]$

Experimental Evaluation

- We implemented **SDa**, **SDd** and **QOS** in C++
 - OpenSSL for cryptographic operations
 - AES-NI enabled
- We compare our schemes with the previous state-of-the-art DSE schemes
 - **HORUS** and **MITRA**
- We measured search time, update time, and communication size
 - Synthetic dataset of **100** to **100M** records
 - Real dataset of **6M** crime incidents in Chicago
- Experiments using r5.8xlarge AWS machines
 - 32-core Intel Xeon 8259CL 2.5GHz processor
 - Running Ubuntu 16.04 LTS, with 256GB RAM, 100GB SSD (GP2), and AES-NI enabled.

Our code is available here: <https://github.com/jgharehchamani/Small-Client-SSE>

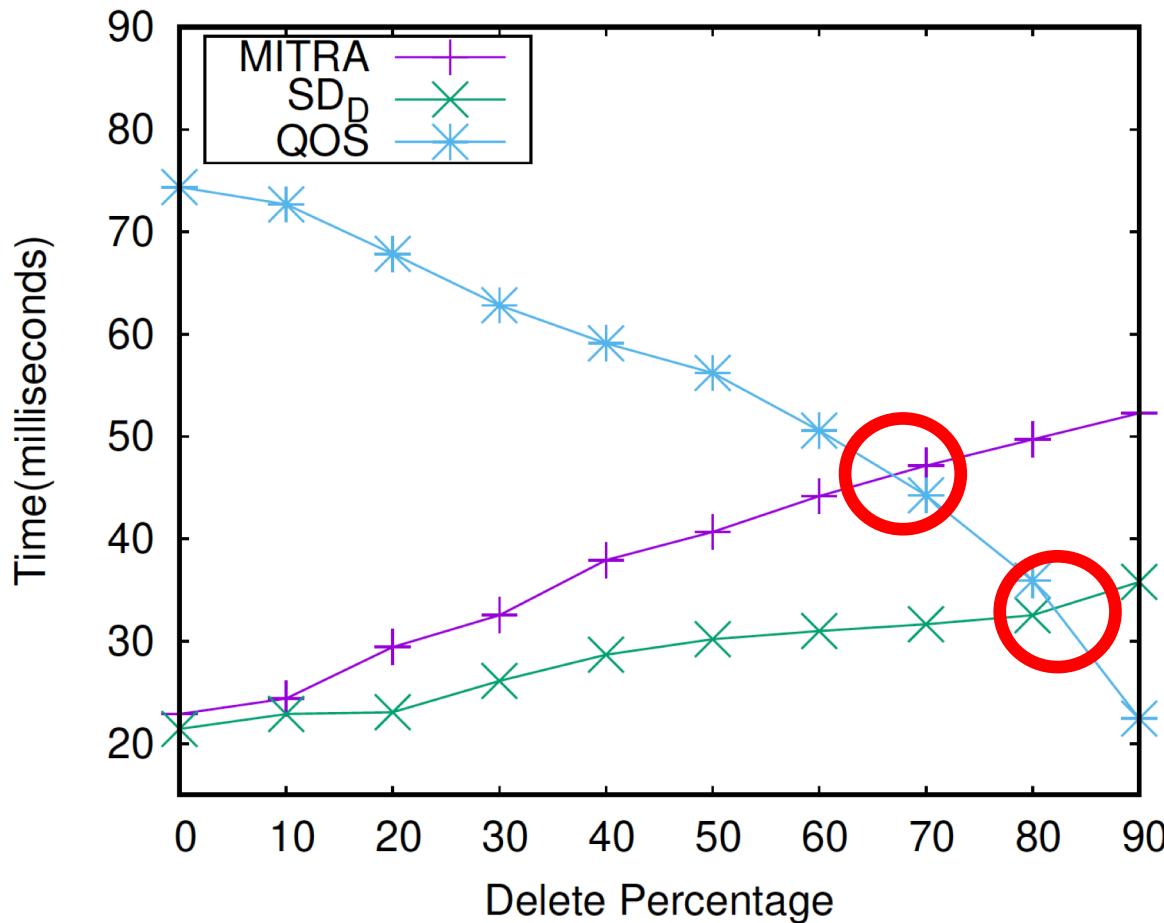
Search Time with 10% Deletions



- Synthetic Dataset **1M records**

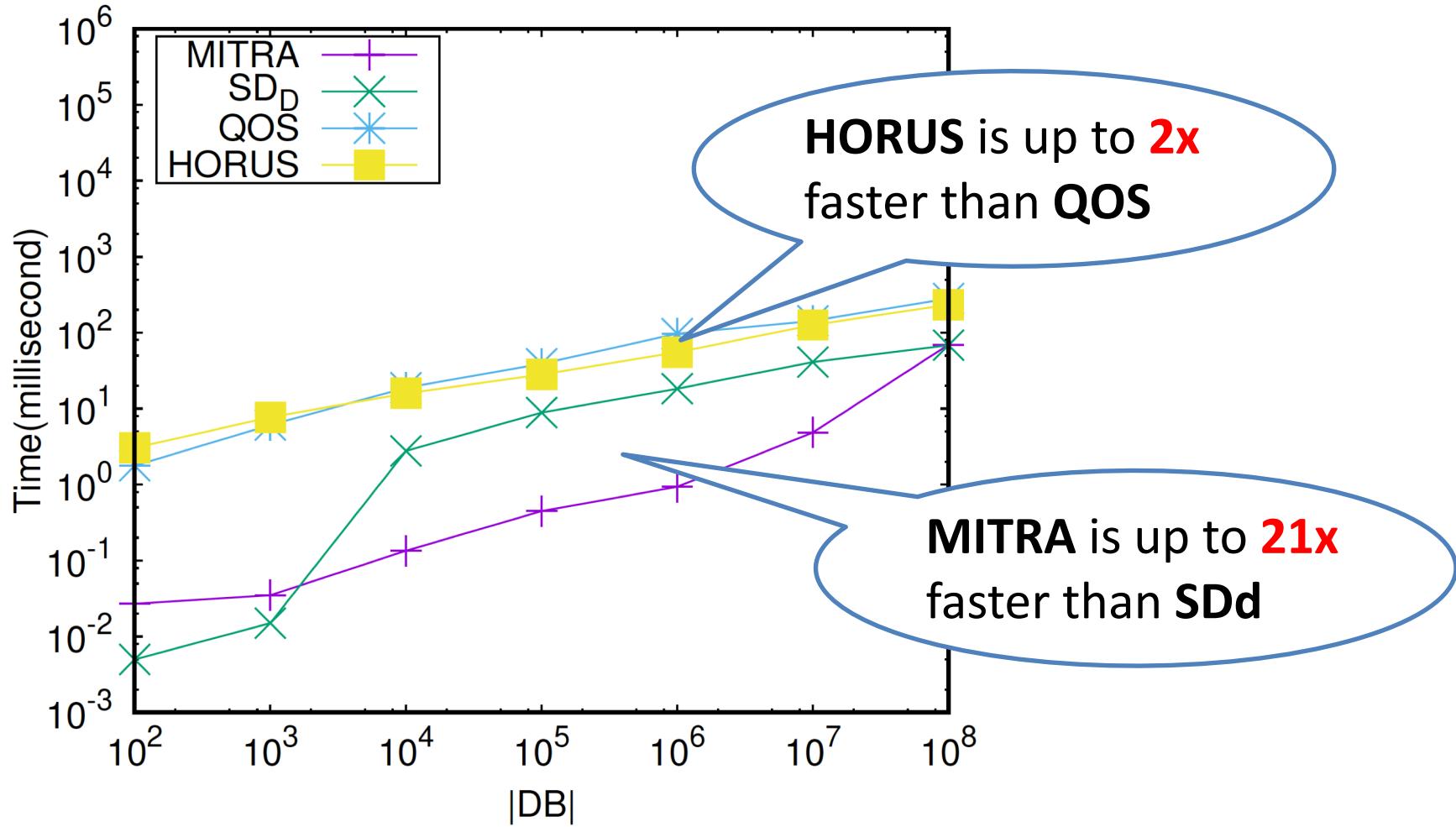
- Result size **100 records**

Search Time with 0-90% Deletion Percentage



- Synthetic Dataset 1M records and $i_w=20K$

Update time



- Synthetic Dataset **100-100M records**

Conclusion

	Search	Update	Search RT	BP-Type
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SDa and SDd

- **20x-85x** faster search time than MITRA
- Minimal client storage & non-interactive search

QOS (for delete intensive query workloads)

- is the state-of-the-art DSE with **quasi-optimal** search time
- **14x-16531x** faster search time than HORUS
- better search time than MITRA and SDd after different deletion ratios between **40%-80%**

Thank You! Questions?



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SDa and SDd

- **20x-85x** faster search time than MITRA
- Minimal client storage & non-interactive search

QOS (for delete intensive query workloads)

- is the state-of-the-art DSE with **quasi-optimal** search time
- **14x-16531x** faster search time than HORUS
- better search time than MITRA and SDd after different deletion ratios between **40%-80%**