



Dynamic Searchable Encryption in Very Large Databases: Data Structures and Implementation

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Searchable Symmetric Encryption (SSE)

■ Definition

- Client encrypts its own data (with its own keys): DB → EDB
- Outsources EDB to a cloud server, keeps a single cryptographic key K
- Later, using K only, performs *keyword-based* searches by sending the cloud *encrypted queries*, and receiving back the *encrypted matching records*

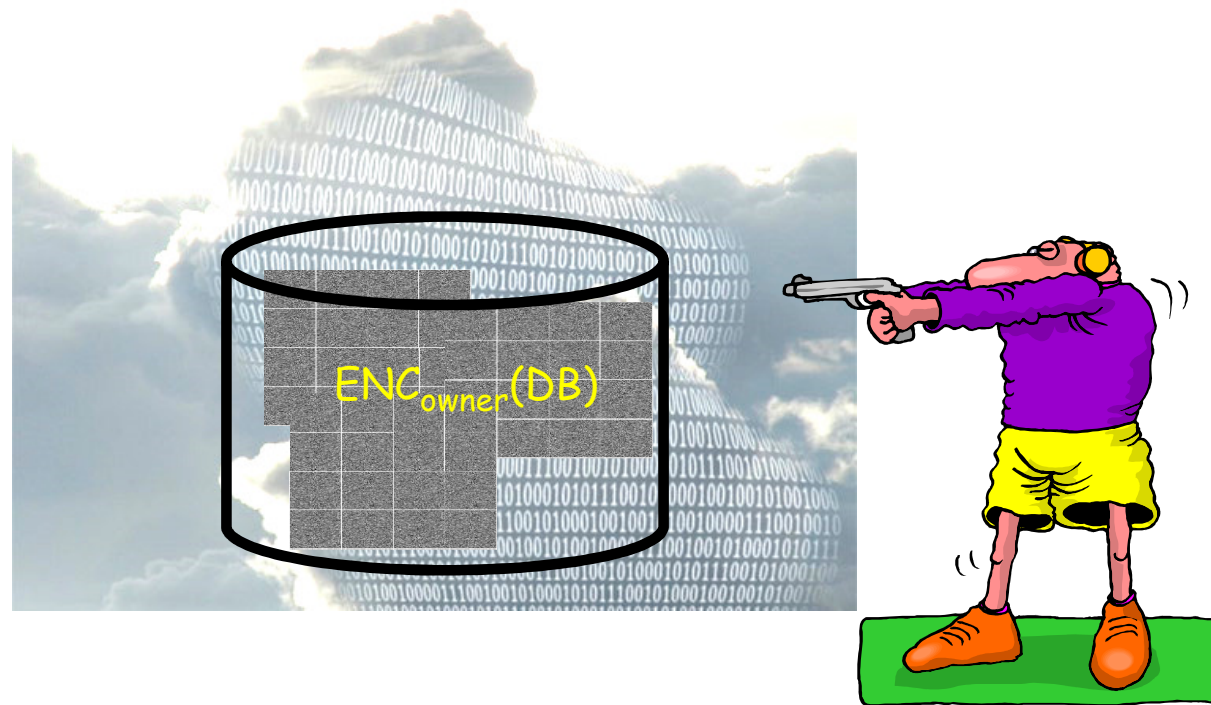
■ Security goal: *Cloud does not learn plaintext data or queries*

- Some forms of statistical leakage allowed: data access patterns (e.g. repeated retrieval, size info), query patterns (e.g., repeated queries), etc.
 - Plaintext data/queries *never directly exposed*, but statistical inference possible
 - Security argued on the basis of **formal leakage profiles** and well defined adv's

■ Application: outsourced private data repositories (email, file system, backup, database, ...)

With SSE...

The cloud cannot disclose your data... *not even at gun point!*

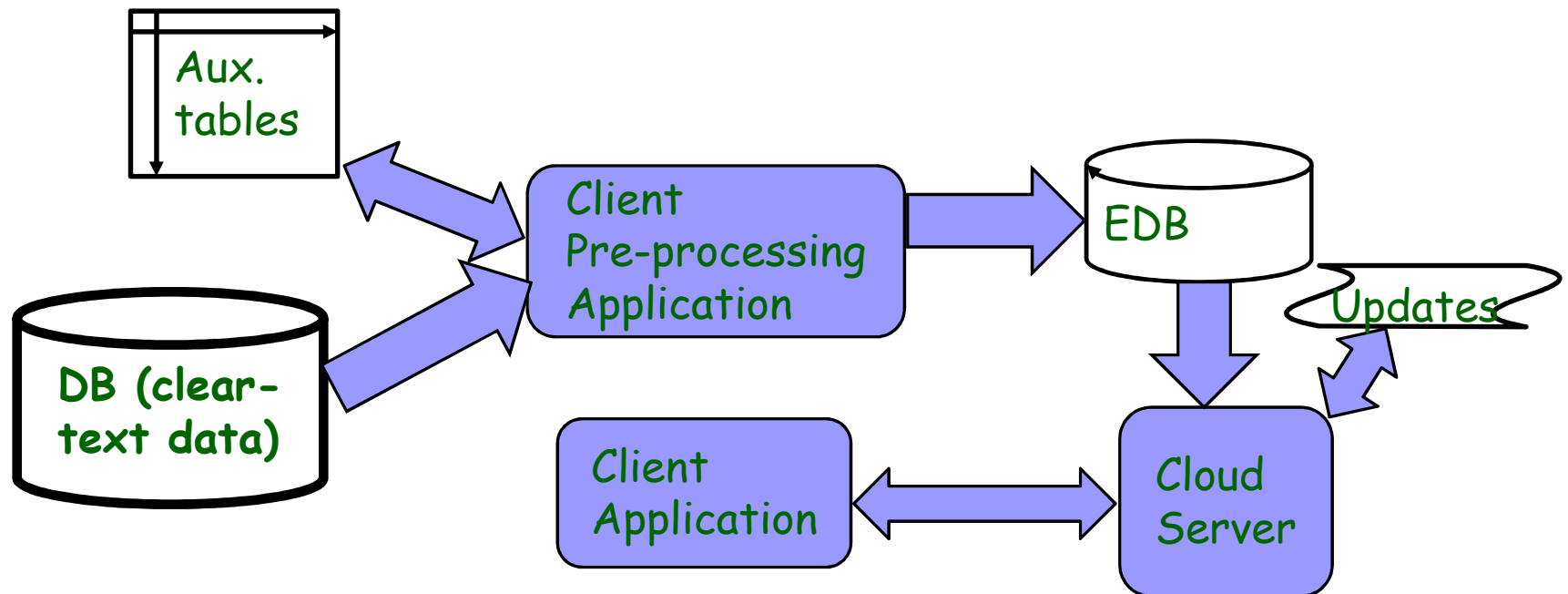




Practical Goals and Trade-offs

- Tens of Billions of distinct (keyword, recId) pairs
 - DB: Relational tables or document collections
 - EDB Workload dominated by searches
 - Encrypted Search Performance should be comparable to Clear-Text Search
 - DB→EDB (pre-processing) done periodically
- Moderate Hardware requirements (set by the funding entity)
 - 4-12 CPU cores, ~100GB RAM, ~10TB additional storage
- Tradeoff: **extensive pre-processing to speed-up encrypted searches**
 - Updates separated from the encrypted database
 - Pre-process to integrate updates or to limit leakage

Prototype



- Carefully designed to scale beyond RAM
 - Big challenge: Security implies maximal randomization yet efficiency calls for maximal "sequentialization" in disk and DB access!!
- Code: 65+k lines of C, lex/yacc & Perl

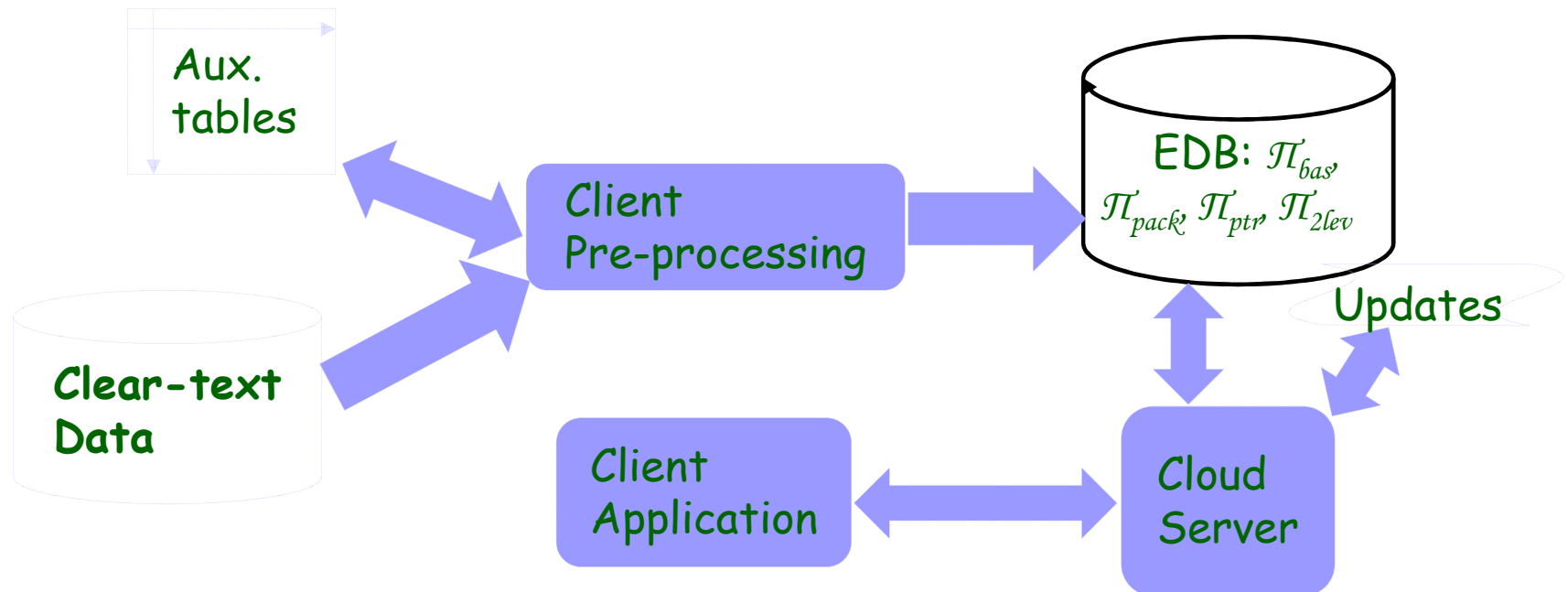
Quantifying Leakage

- Static SSE Scheme for Single Keyword Search (SKS)
 - Setup(DB): Encrypting clear-text data (pre-processing)
 - Search(w): Querying encrypted data
- Static SSE Schemes: $\mathcal{P}_{bas}, \mathcal{P}_{pack}, \mathcal{P}_{ptr}, \mathcal{P}_{2lev}$
 - Leakage functions $\mathcal{L}_{bas}, \mathcal{L}_{pack}, \mathcal{L}_{ptr}, \mathcal{L}_{2lev}$
- Each scheme \mathcal{P} is proven \mathcal{L} -secure against adaptive attacks!!

Quantifying Leakage contd.

- Dynamic SSE Scheme for Single Keyword Search (SKS)
 - Setup(DB): Encrypting clear-text data (pre-processing)
 - Search(w): Querying encrypted data
 - Update: Inserting, Deleting, Modifying records
- Dynamic schemes \mathcal{P}^+ , \mathcal{P}^{dyn} and leakage functions \mathcal{L}^+ , \mathcal{L}^{dyn}
- Each scheme \mathcal{P} is proven \mathcal{L} -secure against adaptive attacks!!
- Integrated formal protocol and system design!
 - \mathcal{P}_{pack} and \mathcal{P}_{2lev} implemented and evaluated!

EDB Data Structures



Big challenge: Security implies maximal randomization yet efficiency calls for maximal “sequentialization” in disk and DB access!

\mathcal{P}_{pack} [Crypto 2013] 100x larger datasets than previous work

\mathcal{P}_{2lev} [NDSS 2014] another 100x over \mathcal{P}_{pack}

EDB as SKS dictionary: $(\text{Enc}_{K_1}(w), \text{EDB}(w)) \forall w$, where $\text{EDB}(w) = \{\text{Enc}_{K_2}(\text{Id}) \mid w \in \text{record}_{\text{Id}}\}$

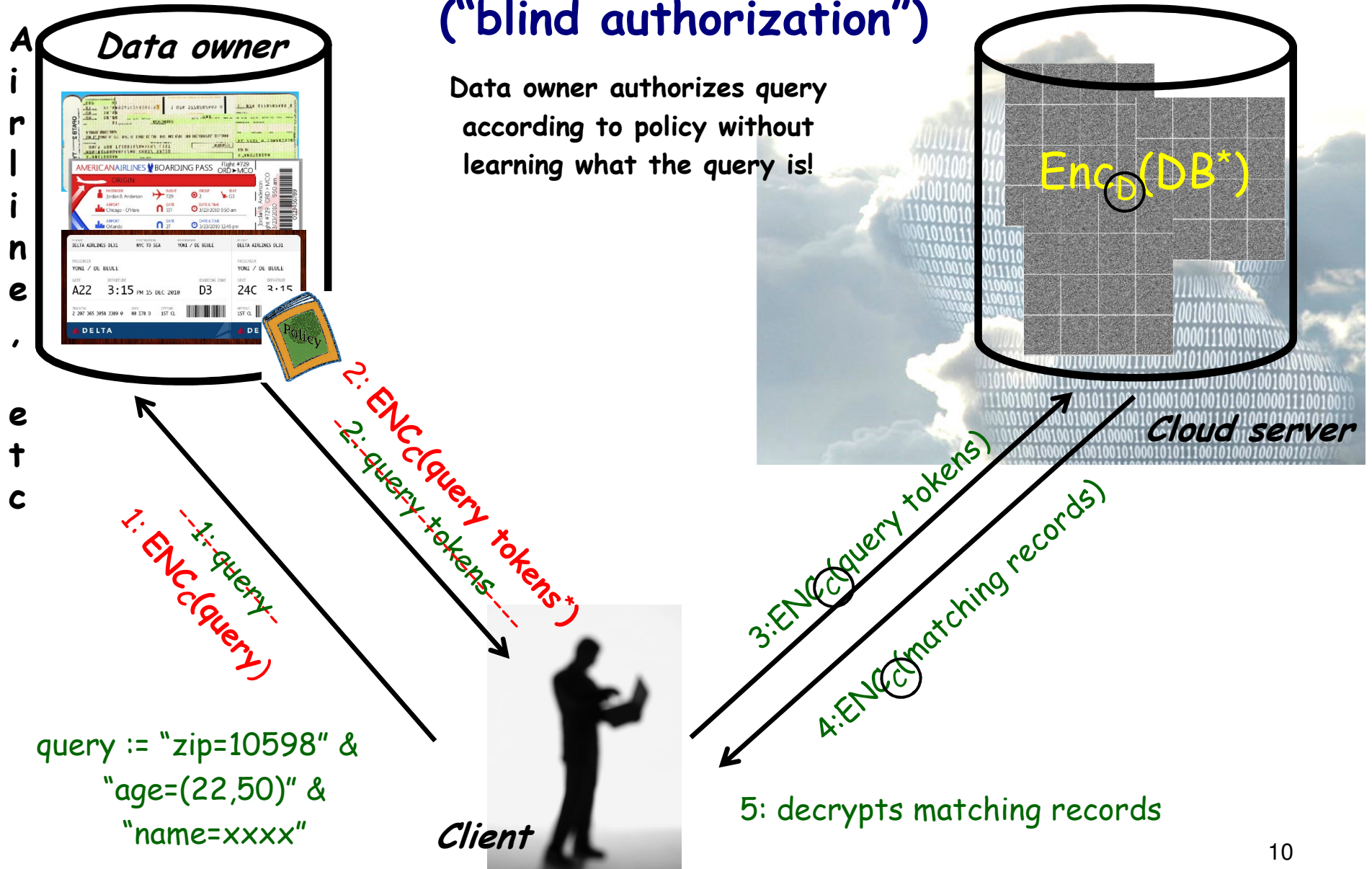
Complex Functional Settings

- Multi-Client SKS SSE: data owner shares its cloud data with friends
 - $EDB_{MC}: (Enc_{K_1}(w), EDB_{MC}(w)), \forall w$ where
$$EDB_{MC}(w) = \{Enc_{K_2}(Id, RDK_{Id}) \mid w \in record_{Id}\} [CCS 2013]$$
- Multi-Client, Conjunctive Search (OXT) in SSE setting
 - $EDB_{OXT}: (Enc_{K_1}(w), EDB_{OXT}(w)), \forall w$ where
$$EDB_{OXT}(w) = \{Enc_{K_2}(Id, RDK_{Id}, 'xind', 'y') \mid w \in record_{Id}\},$$
'xind' and 'y' are required for conjunctive queries [Crypto 2013]
- Outsourced Symmetric PIR: data owner authorizes clients to perform queries (policy)... **without learning the search terms she authorizes**
 - Data owner is malicious but she does not collude with Cloud server
 - 'Data owner' - 'Cloud server' separation crucial to avoid PIR cost.

Outsourced Symmetric PIR Setting

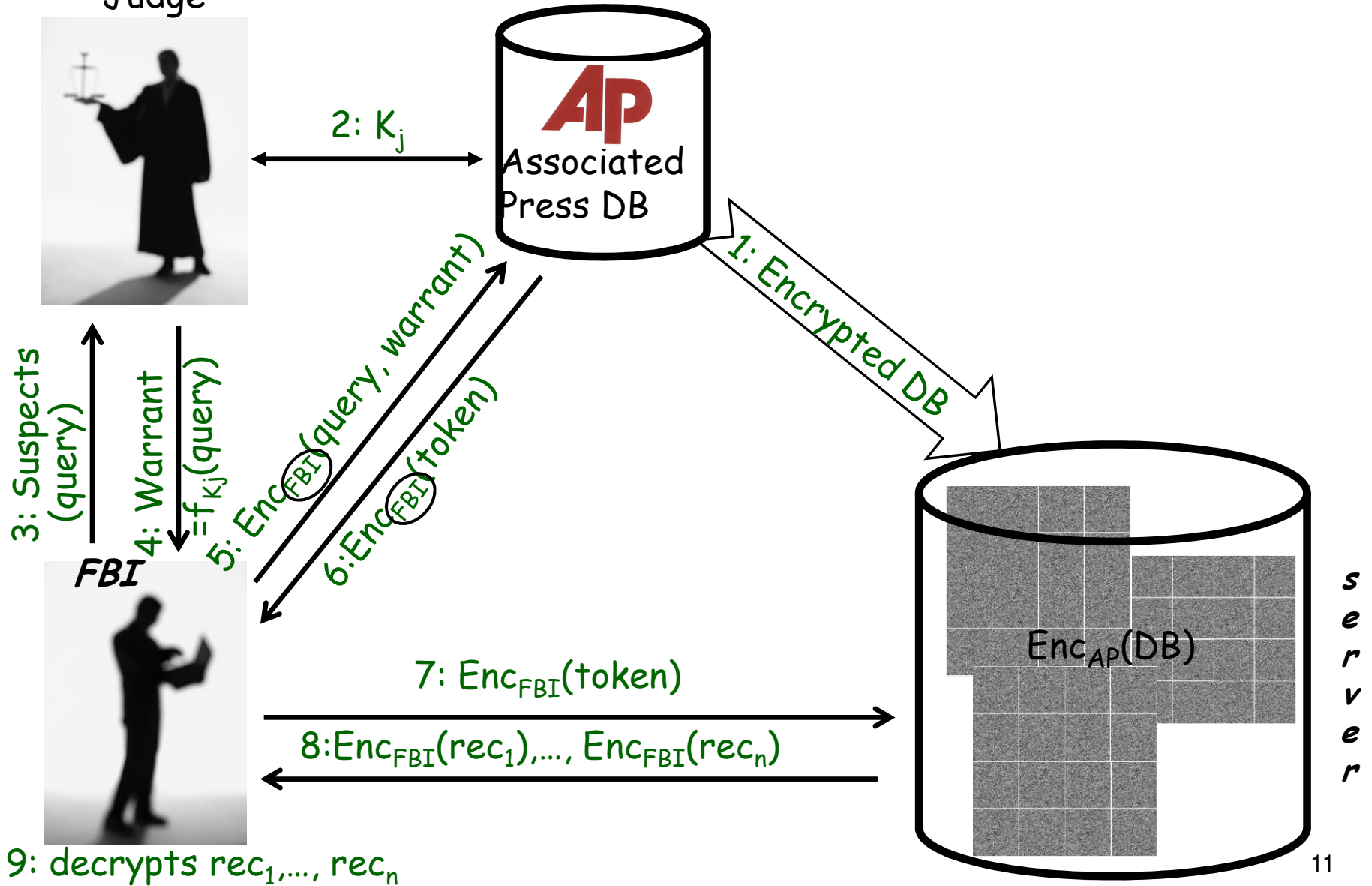
("blind authorization")

Data owner authorizes query according to policy without learning what the query is!



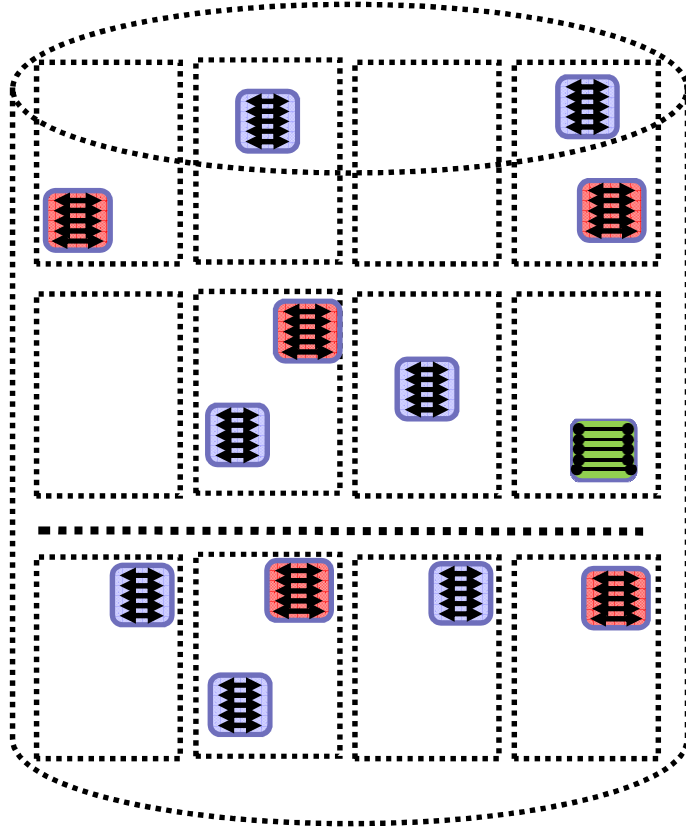
OSPIR w/ Warrant-based Authorization

Judge



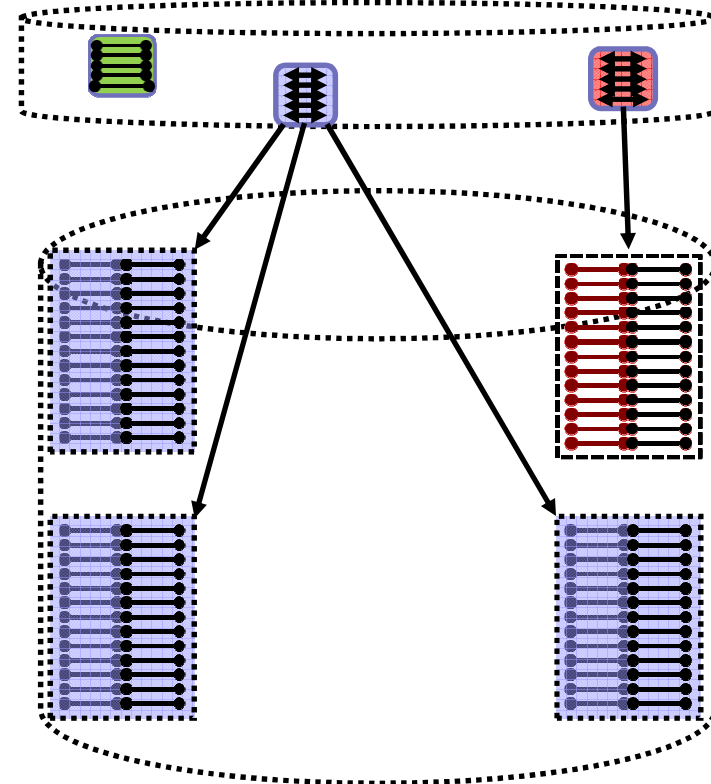
Faster Pre-Processing and Better Goodput

\mathcal{H}_{pack} : Bucket (Paged) Hash (PH)



- Low storage utilization (~60%)
- Cuckoo Hash fix (~90% util): sensitive to insertion history
- Low goodput

\mathcal{H}_{2lev} : Two Levels (2L)

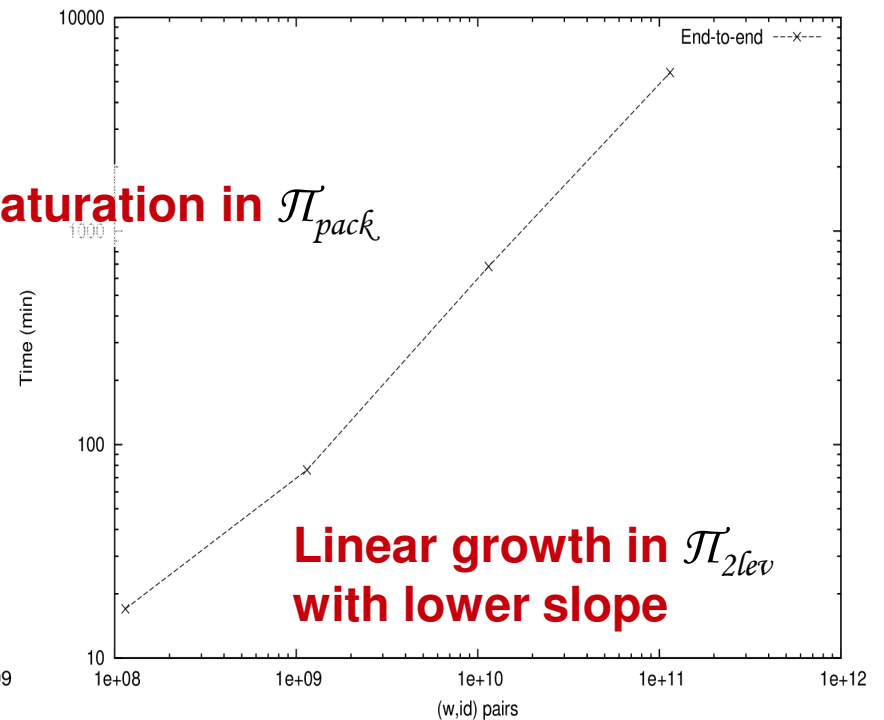
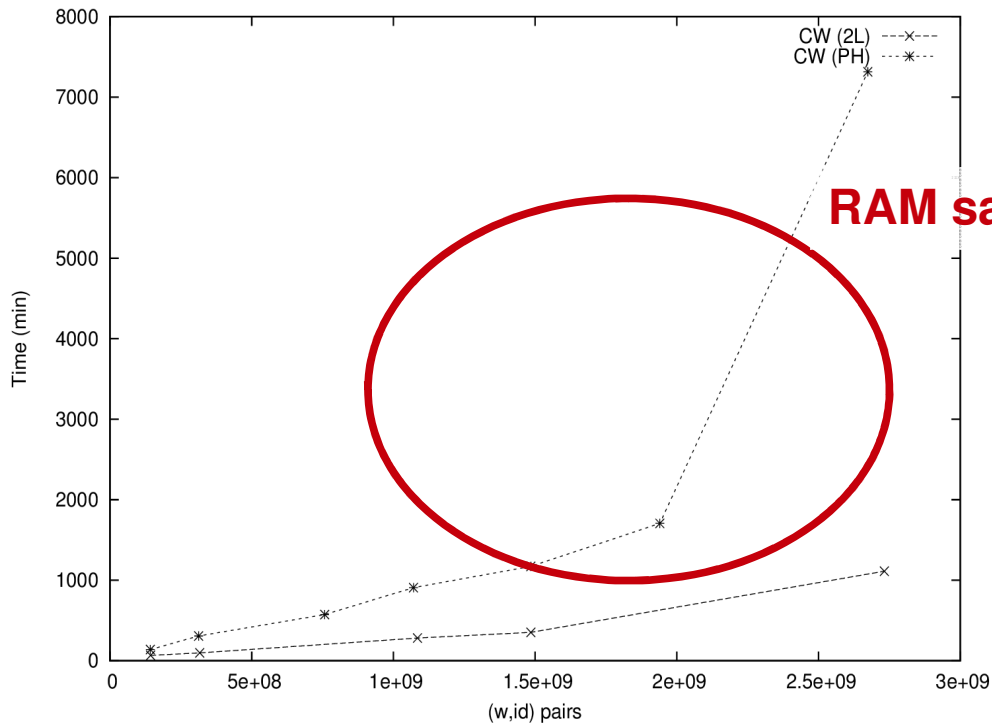


- Multi-modal keyword distribution
- Good storage utilization (92%)
- High goodput.

Pre-processing Scalability

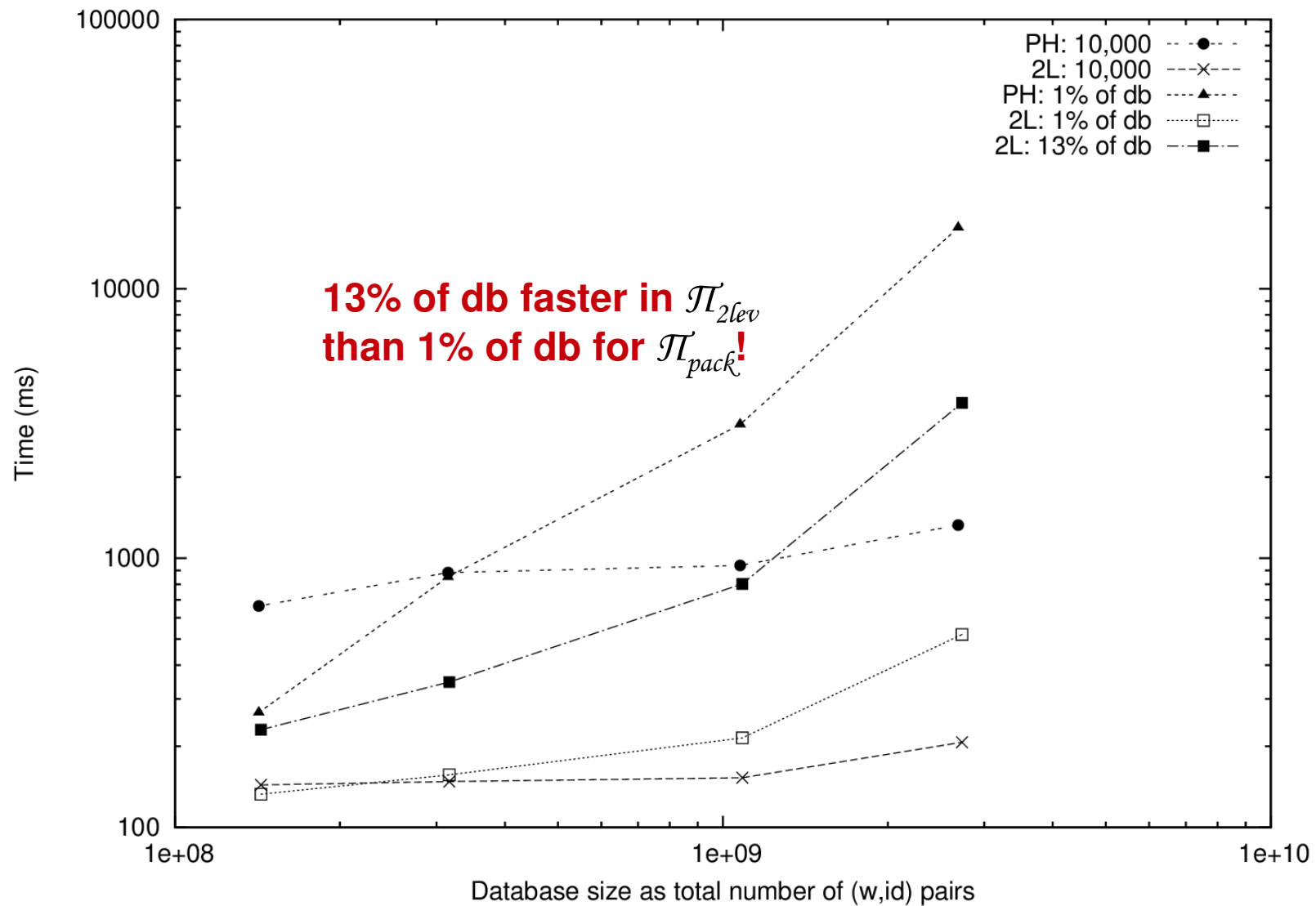
\mathcal{P}_{2lev} vs. \mathcal{P}_{pack} on ClueWeb (OXT)

\mathcal{P}_{2lev} on Census Data (SKS)



*ClueWeb: Subsets of ClueWeb09 data set, crawled web-pages including wikipedia.
Census Data: Lincoln Lab's database.*

SKS Query Scalability: Π_{2lev} vs. Π_{pack}



Acknowledgements and Future Work



- Supported by the Intelligence Advanced Research Projects Activity (IARPA) under the Security and Privacy Assurance Research (SPAR) program
- In the News: President Obama announcing plans for moving Telephone Data away from NSA (speech Jan 17th 2014)

The review group recommended that our current approach be replaced by one in which the providers or a third party retain the bulk records, with government accessing information as needed. Both of these options pose difficult problems. [...] During the review process, some suggested that we may also be able to preserve the capabilities we need through a combination of existing authorities, better information sharing, and recent technological advances. But more work needs to be done to determine exactly how this system might work.

IARPA SPAR Program



Thanks!