

Reconstructing Encrypted Data Using Range Query Leakage

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ePrint 2017/701, to appear S&P 2018.

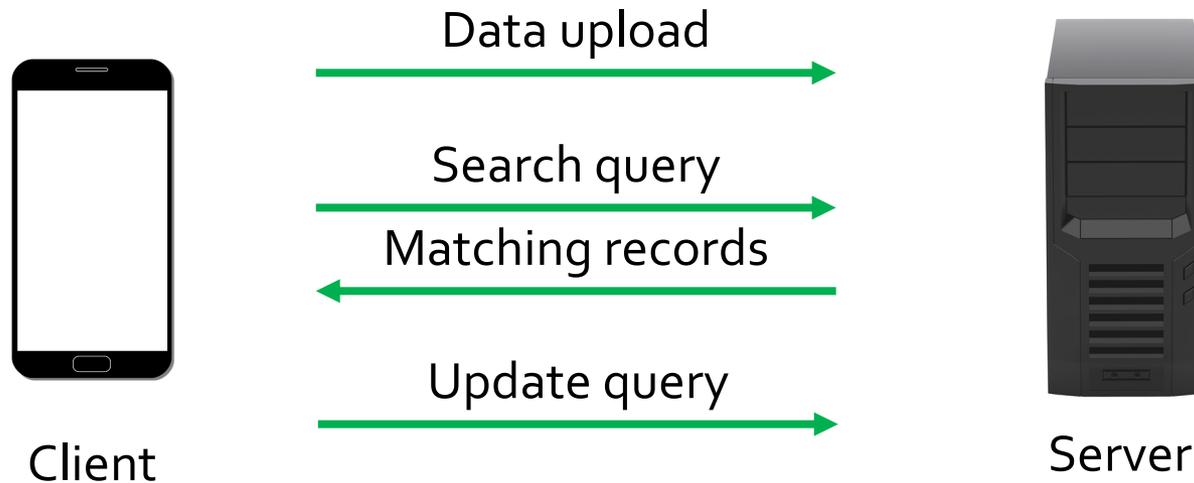
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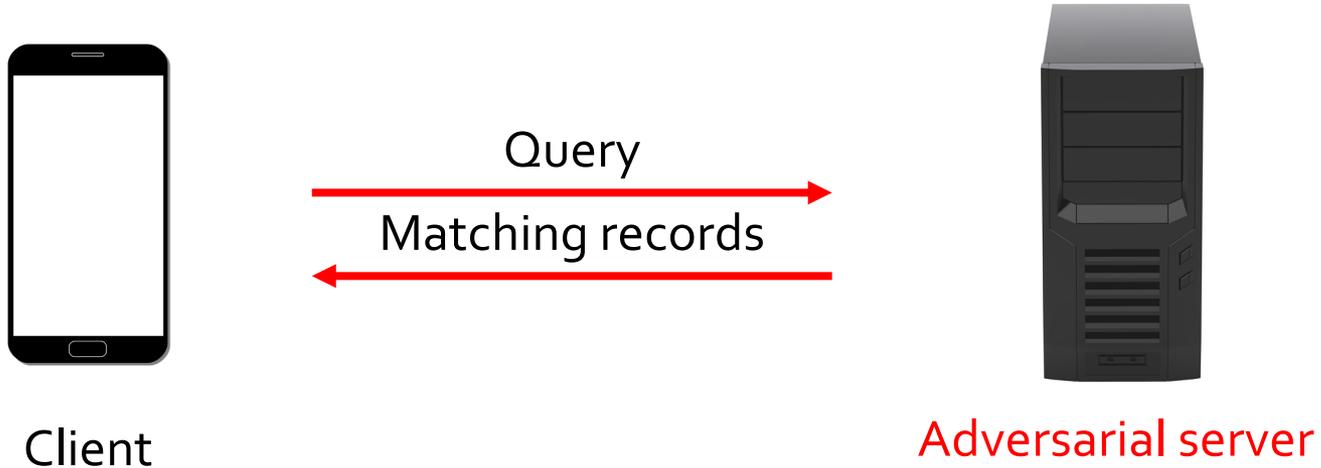
Workshop IoT+Cloud, Bochum, 7 Nov 2017.

Outsourcing Data to the Cloud



- For **encrypted database management systems**:
 - Data = collection of records in a database (e.g. health records).
 - Query examples =
 - Find records with a given value (e.g. patients aged 57).
 - Find records within a given range (e.g. patients aged 55 to 65).
 - ...

Security of Data Outsourcing Solutions



- **Adversaries:**
 - **Snapshot** adversary = breaks into server, gets snapshot of memory.
 - **Persistent** adversary = corrupts the server for a period of time. Sees all communication transcripts. Can be server itself.
- **Security goal = privacy:**

Adversary learns as little as possible about the client's data and queries.

State of the Art

- **No perfect solution.**

Every solution is a trade-off between **functionality** and **security**.

- **Huge amount of literature.**

[AKSX04], [BCLO09], [PKV+14] , [BLR+15], [NKW15], [K15],
[CLWW16], [KKNO16] , [RACY16], [LW16] ...

- **A few “complete” solutions:**

Mylar (for web apps)

CryptDB (handles most of SQL)

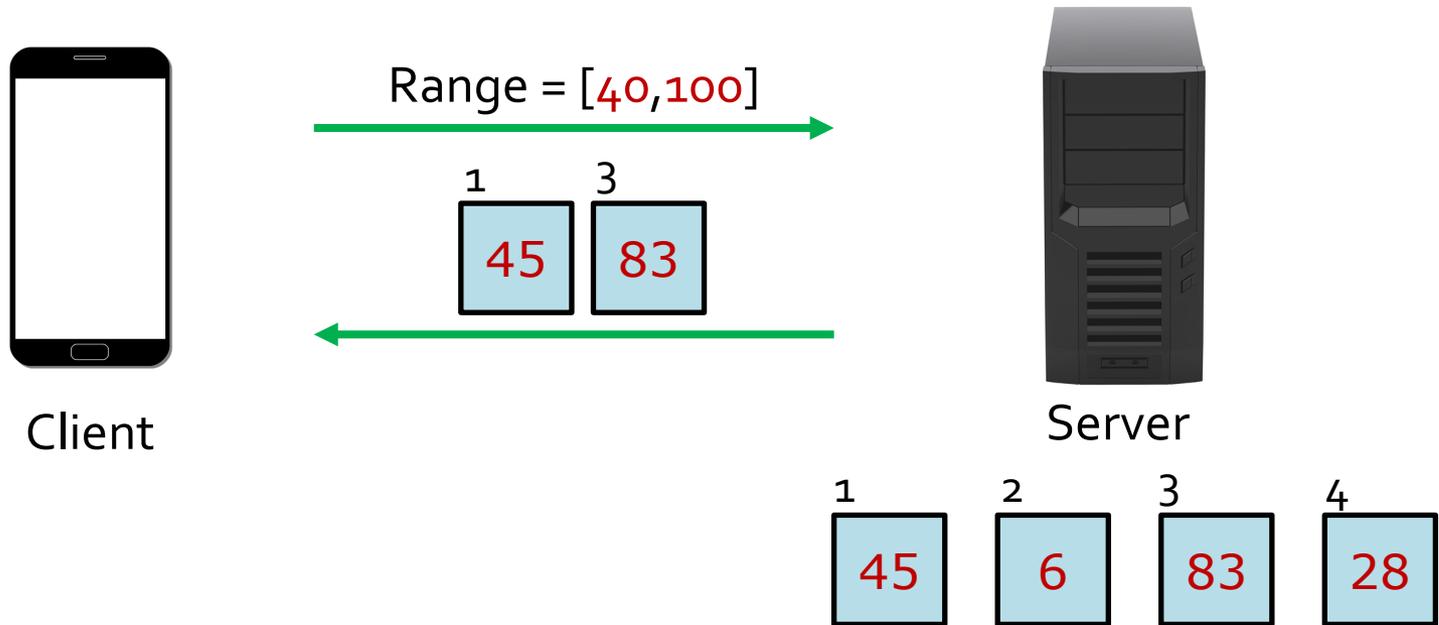


! *Controversial!*

→ Cipherbase (Microsoft), Encrypted BigQuery (Google), ...

- **Very active area of research.**

Setting for this Talk: Schemes Supporting Range Queries



- All known schemes leak set of matching records = **Access Pattern**.
OPE, ORE schemes, POPE, [HK16], Blind seer, [Lu12], [FJKNRS15],...
- Some schemes also leak # records below queried range endpoints = **rank**.
FH-OPE, Lewi-Wu, Arx, Cipherbase, EncKV,...

Exploiting leakage

- Most schemes prove that nothing more leaks than their leakage model allows.
- For example, leakage = **access pattern**, or **access pattern + rank**.
- *What can we really learn from this leakage?*
- **Our goal:** full reconstruction = recover the exact value for every record.
- **[KKNO16]:** $O(N^2 \log N)$ queries suffice for full reconstruction using only access pattern leakage!
 - where N is the number of possible values (e.g. 125 for age in years).

Assumptions for our Analysis

1. Data is **dense**: all values appear in at least one record.
2. Queries are **uniformly distributed**.

Our algorithms don't actually care though – the assumption is for computing data upper bounds.

Our Main Results

- **Full reconstruction** with $O(N \cdot \log N)$ queries from **access pattern**
 - in fact, $N \cdot (3 + \log N)$.
- **Approximate reconstruction** with relative accuracy ε with $O(N \cdot (\log 1/\varepsilon))$ queries.
- **Approximate reconstruction** using an *auxiliary distribution* and **rank** leakage.
 - more efficient in practice, evaluation via simulation.



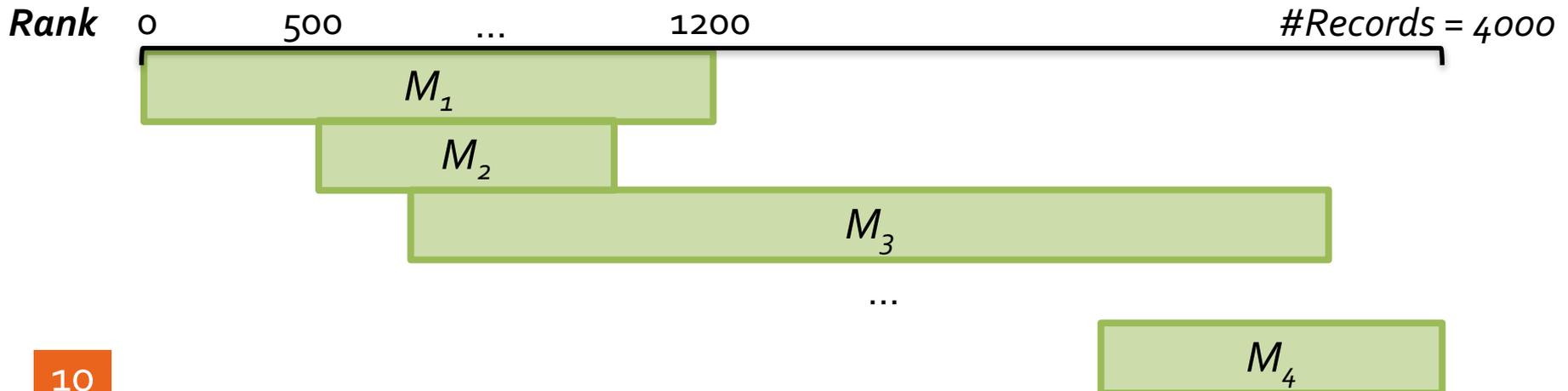
Attack 1: Full Reconstruction

Full Reconstruction with Rank Leakage

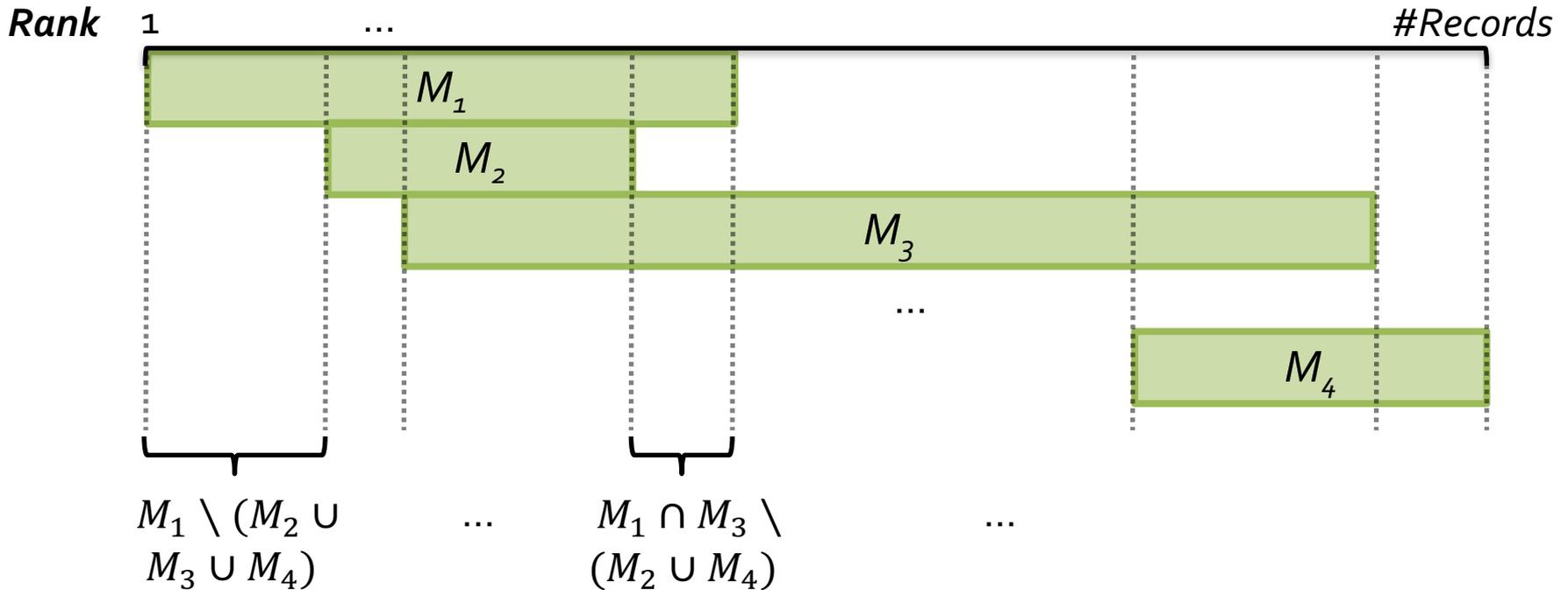
- Adversary is observing query leakage...

Hidden	Leaked		
Query [x,y]	a = rank(x-1)	b = rank(y)	Matching IDs
[1,18]	0	1200	M_1
[2,10]	500	800	M_2
[7,98]	600	3000	M_3
[55,125]	2000	4000	M_4

(Reordered for convenience)



Full Reconstruction with Rank Leakage



- Partition records into smallest possible sets using access pattern leakage.
- If this partitions records into N sets, **win!** Just match minimal sets with values.

Full Reconstruction with Rank Leakage

- Expected number of queries **sufficient** for **full reconstruction** is at most:

$$N \cdot (2 + \log N) \quad \text{for } N \geq 27.$$

Essentially a coupon collector's problem.

- Expected number of **necessary** queries is at least:

$$\frac{1}{2} \cdot N \cdot \log N - O(N)$$

for *any* algorithm.

- This algorithm is “**data-optimal**”, i.e. it fails iff full reconstruction is impossible for *any* algorithm given the input data.

Full Reconstruction **without** Rank Leakage

- **Very generic setting:** use only **access pattern** leakage.
- **Partition** (as before), then **sort**.
- Expected number of **sufficient** queries is at most:
$$N \cdot (3 + \log N) \quad \text{for } N \geq 26$$
 - i.e. sorting step is very cheap in terms of data.
- Expected number of **necessary** queries is at least:
$$1/2 \cdot N \cdot \log N - O(N)$$

for *any* algorithm.
- Still **data-optimal!**



Attack 2: Reconstruction with Auxiliary Data

Reconstruction with Auxiliary Data and Rank Leakage

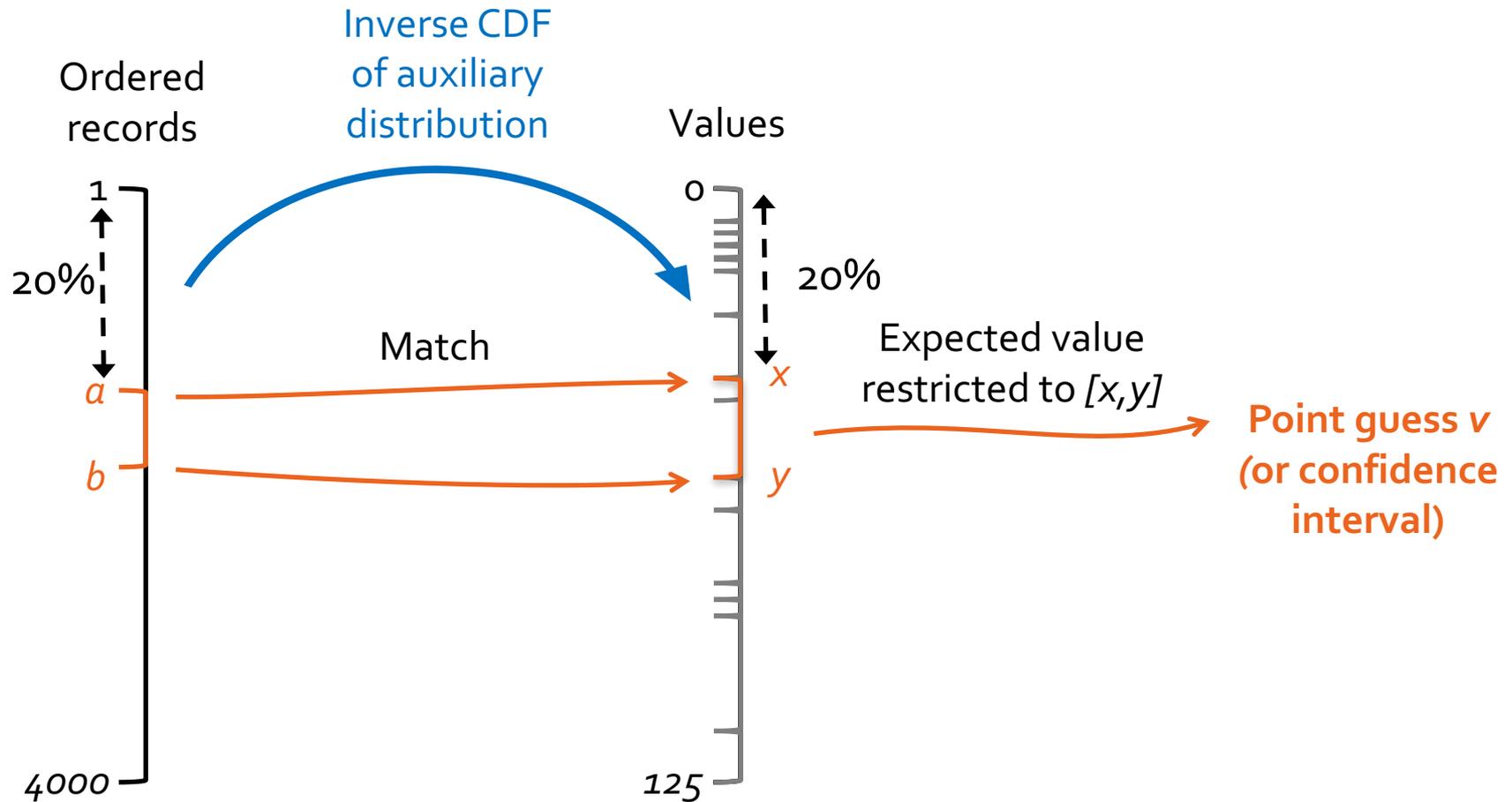
- As before, queries have ranges chosen uniformly at random.
- Assume **access pattern** and **rank** are leaked.
- We now also assume that an **approximation to the distribution on values** is known.

“Auxiliary distribution”.

From aggregate data, or from another reference source.

- We show experimentally that, under these assumptions, **far fewer queries** are needed.

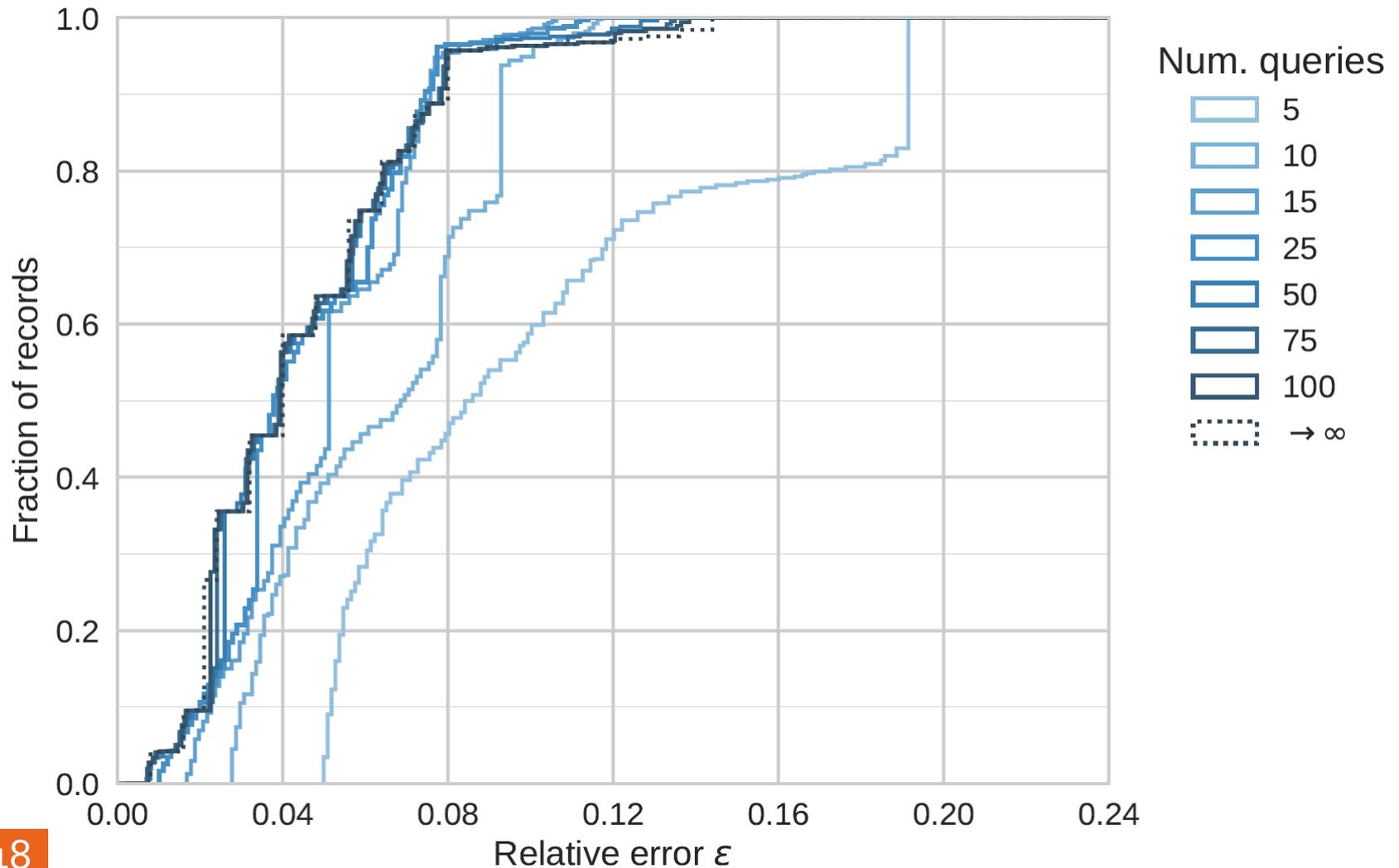
Auxiliary Data Attack: Estimating Step



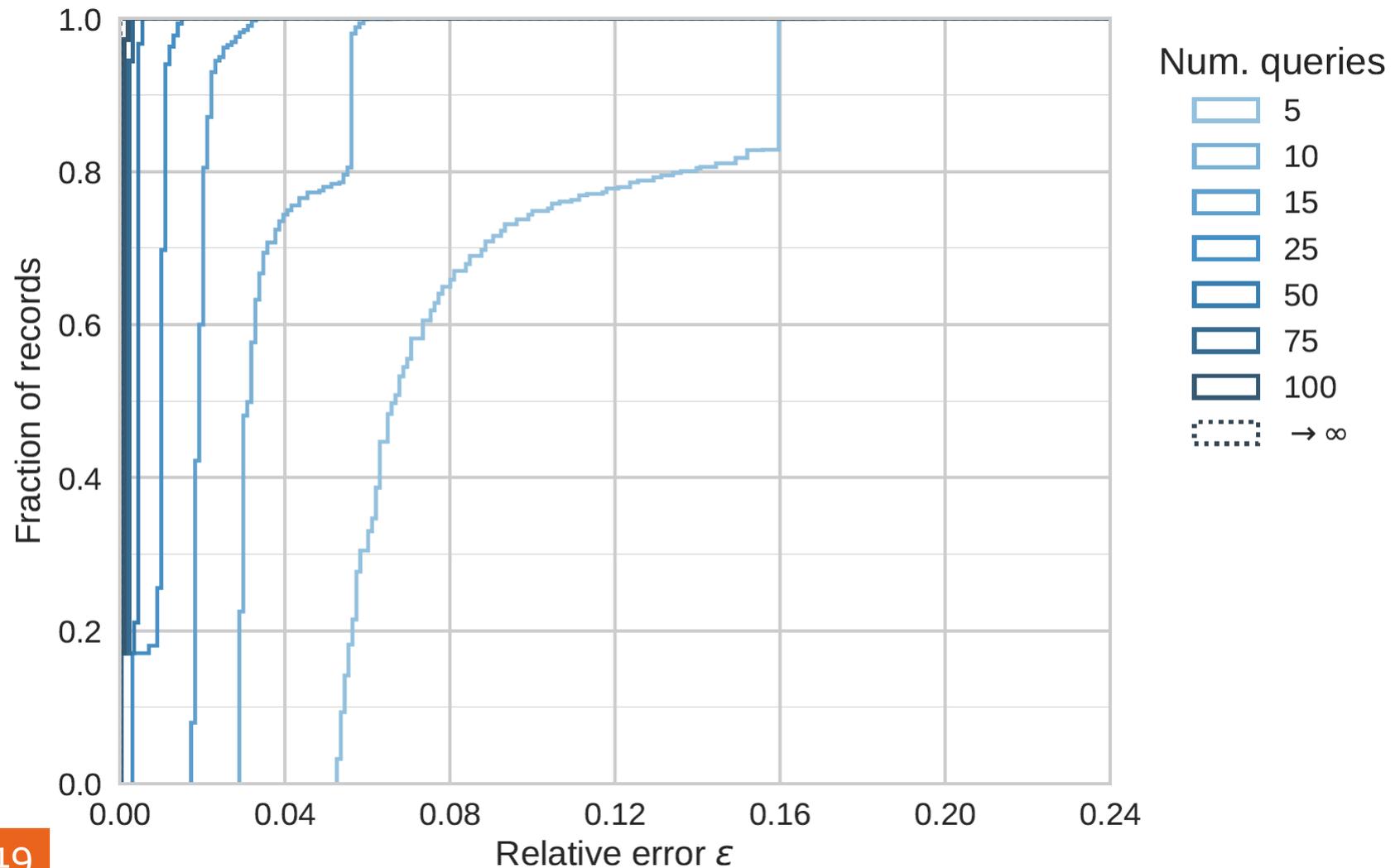
Auxiliary Data Attack: Experimental Evaluation

- Ages, $N = 125$ (0 to 124).
- Health records from US hospitals (NIS HCUP 2009).
- **Target:** age of individual hospitals' records.
- **Auxiliary data:** aggregate of 200 hospitals' records.
- **Measure of success:** proportion of records with value guessed within ϵ .

Auxiliary Data Attack: Results for Typical Target Hospital



Auxiliary Data Attack: Results with Perfect Auxiliary Distribution





Summary and Conclusions

Summary of the attacks

- Our results : **full reconstruction** in $\approx N \log N$ queries with only **access pattern!**
Efficient, data-optimal algorithms + matching lower bound.

Attack	Req'd leakage	Other req'ts	Suff. # queries
KKNO₁₆	AP	Density	$O(N^2 \log N)$
Full	AP + rank	Density	$N \cdot (\log N + 2)$
	AP	Density	$N \cdot (\log N + 3)$
ϵ-approximate	AP	Density	$5/4 N \cdot (\log 1/\epsilon) + O(N)$
Auxiliary	AP + rank	Auxiliary dist.	Experimental

- For $N = 125$, about 800 queries suffice for **full reconstruction!**
- If an auxiliary distribution + **rank** leakage is available, after only 25 queries, 55% of records can be reconstructed to within 5 years!

Conclusions

- Many clever schemes have been designed, enabling range queries on encrypted data.

OPE, ORE schemes, POPE, [HK16], Blind seer, [Lu12], [FJKNRS15], FH-OPE, Lewi-Wu, Arx, Cipherbase, EncKV,...

- Second-generation schemes **defeat the snapshot adversary** (with caveats).
- But as our attacks show, **no known scheme offers meaningful privacy vs. a persistent adversary** (including server itself).

In realistic settings, $N \log(N)$ queries suffice; even less if auxiliary distribution + rank leakage is known.

- More research needed!