

Hiding the access
pattern is not enough:

Exploiting search pattern leakage in Searchable Encryption

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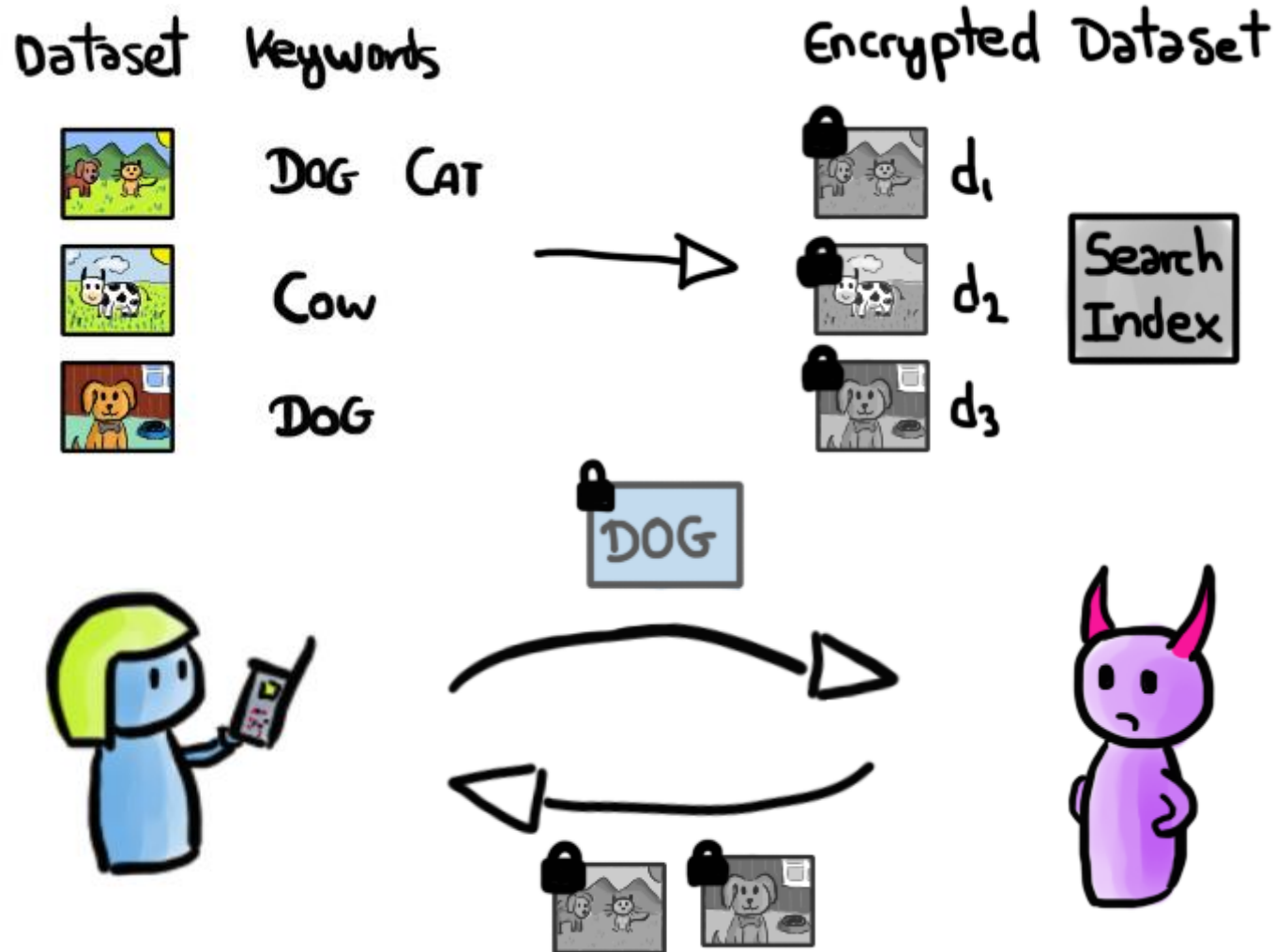
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University of Waterloo

CrySP

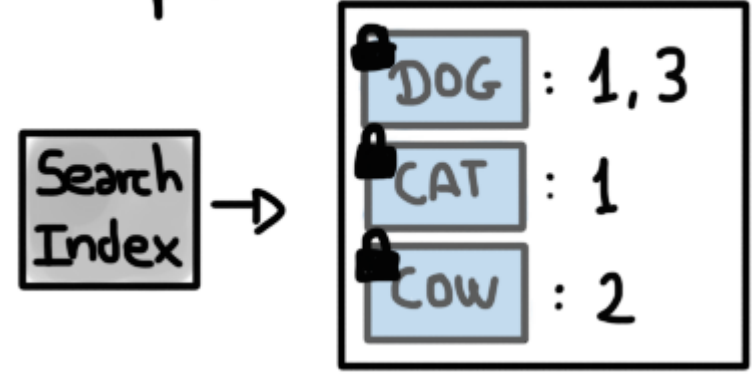


Overview: searchable encryption



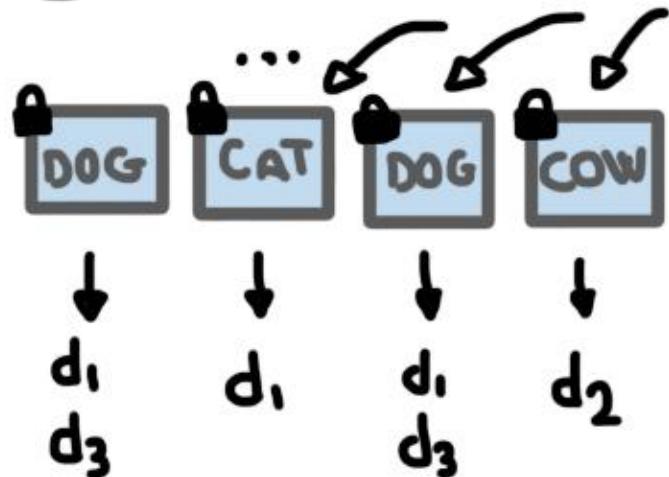
Searchable encryption allows a user to send her encrypted database to a server while still being able to perform secure searches over it.

Example:



Service provider = (Passive) Adversary

Leakage

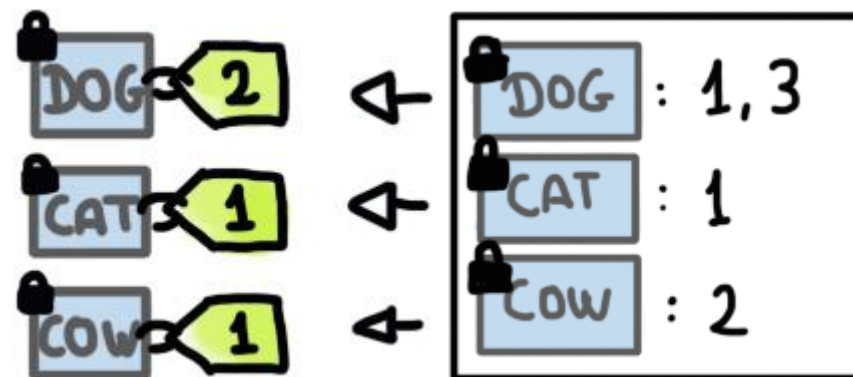
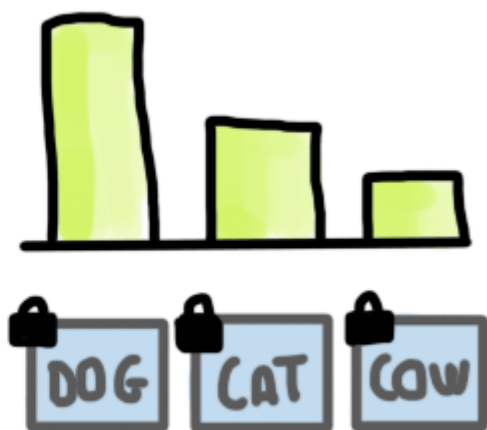


Efficient searchable encryption schemes leak the **search pattern** (which can be used to compute the frequency of each query token) and the **access pattern** (which reveals how many documents match the query).

Access pattern

Keyword volume

Search pattern
↓
Query frequency

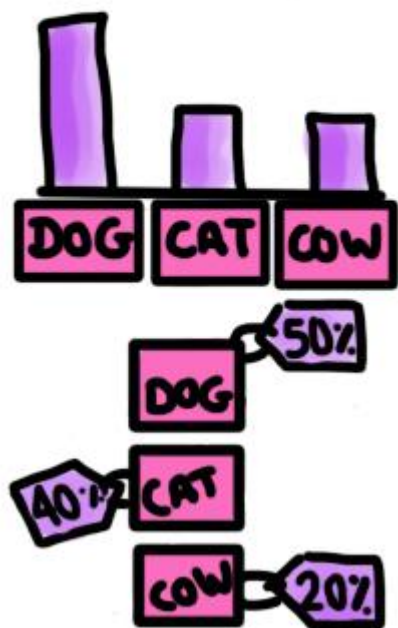


New attack: SAP

Search & Access Pattern-based attack

We propose a new attack (SAP) that uses both search and access pattern leakage, as well as auxiliary information (which is not necessarily ground-truth information).

Aux. Info.



Keywords

DOG

CAT

COW

Query Tokens

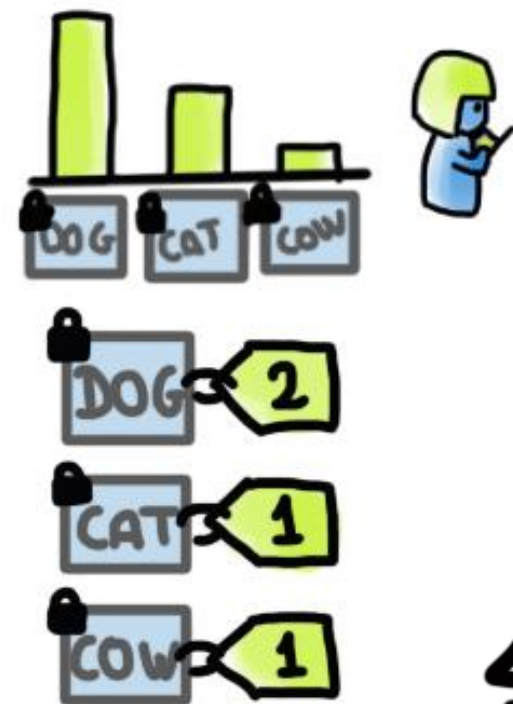
DOG

COW

CAT

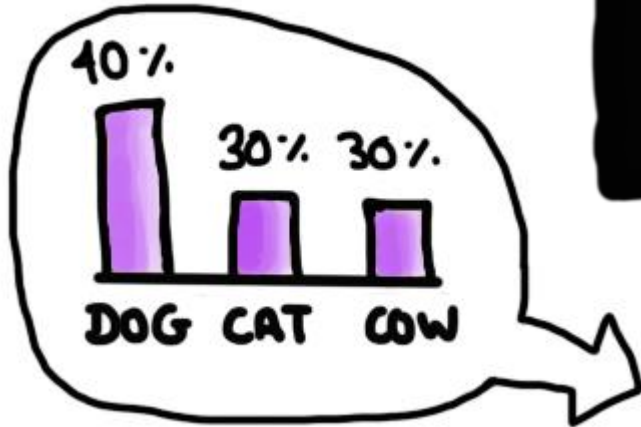
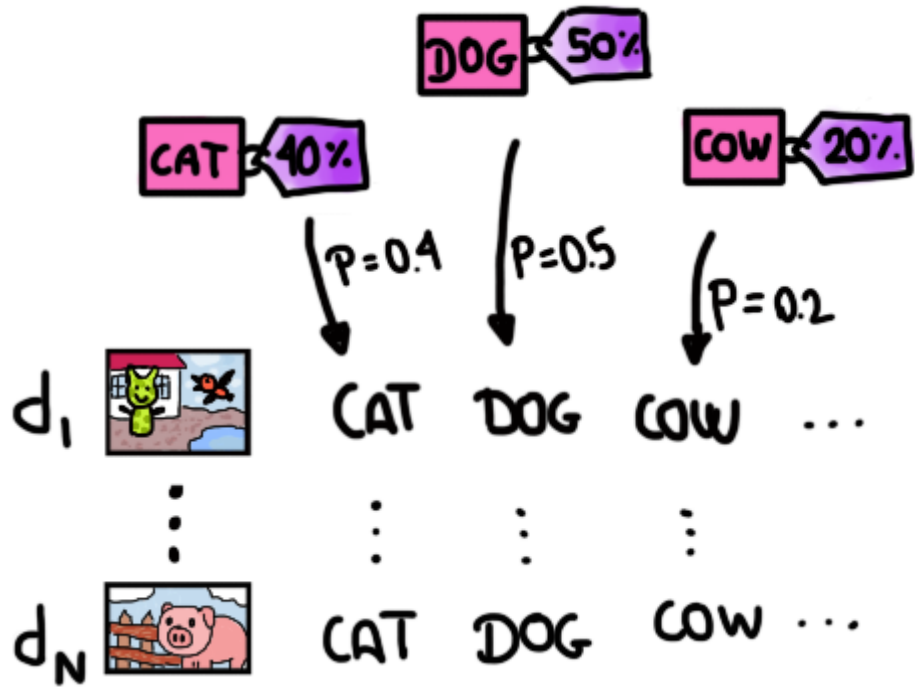


Observations

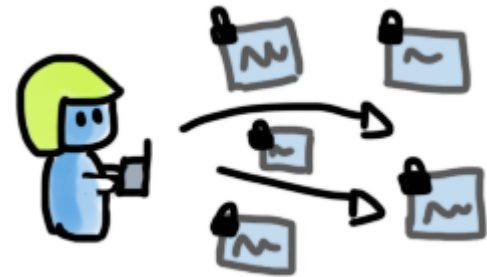


Modelling Observations

To derive the attack, we build a mathematical model of the response volume and frequency of each query, based on the auxiliary information.



η queries



$$(\# \text{DOG}, \# \text{CAT}, \# \text{COW}) \sim \text{Multi}(\eta, (40\%, 30\%, 30\%))$$



$\# \text{ docs} \sim \text{Bino}(N, p = 50\%)$

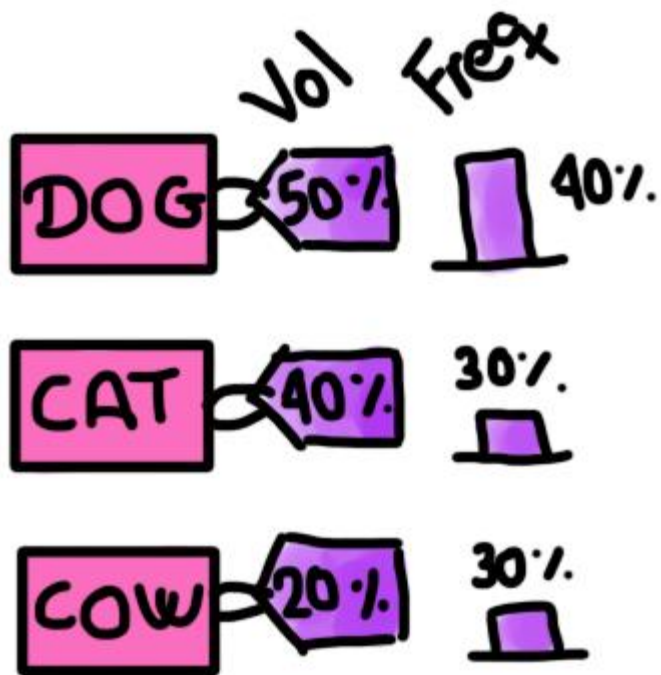


SAP uses ML (Maximum Likelihood)

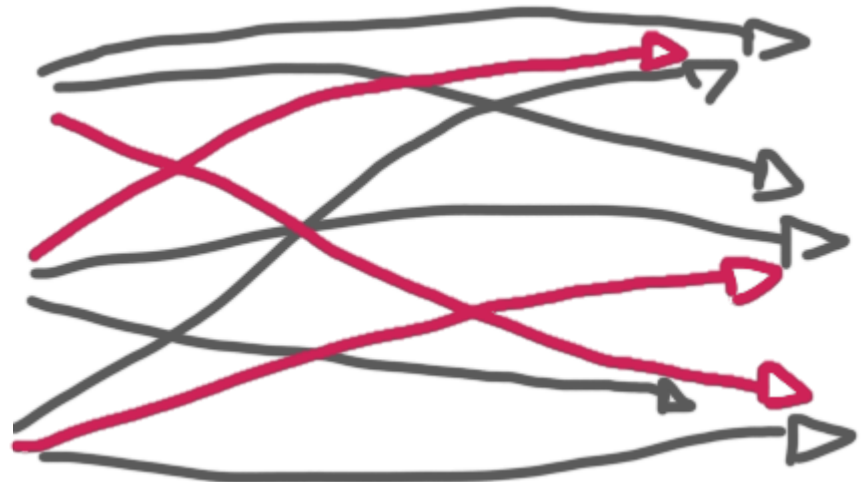
$$\hat{P} = \underset{P \in \mathcal{P}}{\operatorname{argmax}} \Pr(\text{Obs} | \text{Aux}, P)$$



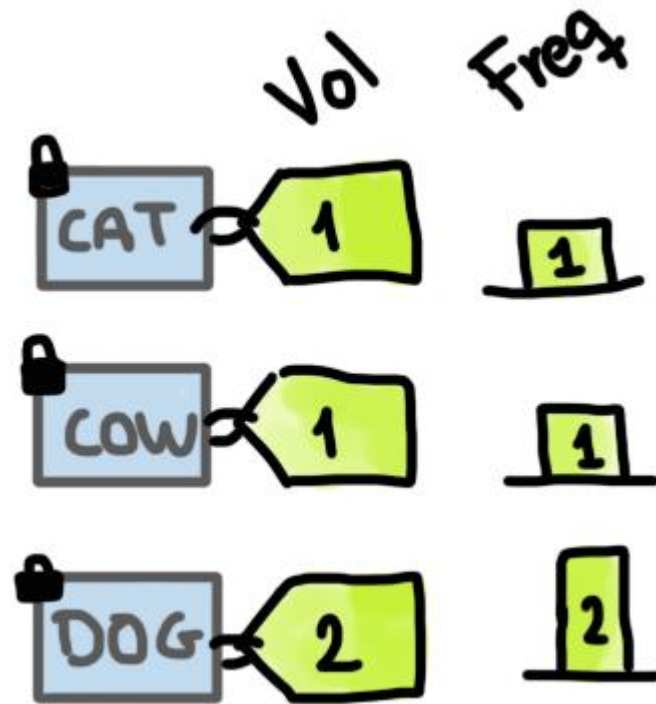
The attack finds the **maximum likelihood** matching of keywords to query tokens given the previous mathematical model. It uses the Hungarian Algorithm to find the optimal matching.



$$\Pr(\text{1}, \text{1} | \text{50\%}, \text{40\%})$$



Hungarian Algorithm



SAP vs. Defenses

SAP is easy to adapt against different volume-hiding defenses (padding) by just taking the defense into account in the mathematical model.

#docs with "DOG"



No defense:

$$? \sim \text{Bino}(N, 50\%)$$

CLRZ $? \sim \text{Bino}(N, 50\% \cdot \text{TPR} + (1 - 50\%) \cdot \text{FPR})$

PPYY $? \sim \text{Bino}(N, 50\%) + \text{Lap}(b) + c$

SEAL $? \sim \text{Pad}_x(\text{Bino}(N, 50\%))$



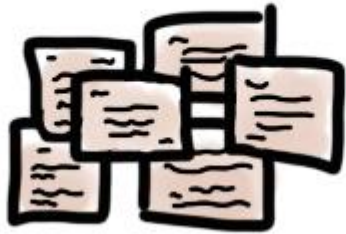
CLRZ: Chen et al. "Differentially private access patterns for searchable symmetric encryption". INFOCOM'18

PPYY: Patel et al. "Mitigating leakage in secure cloud-hosted data structures". CCS'19

SEAL: Demertzis et al. "SEAL: Attack mitigation for encrypted databases via adjustable leakage". USENIX'20

Evaluation:

Real datasets



50% →

Aux Info.

50% →

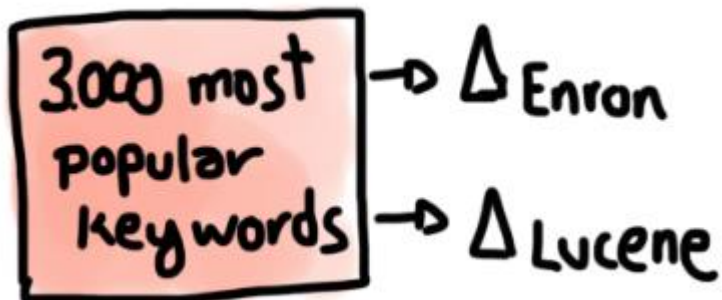
Actual dataset

We evaluate SAP using real datasets (Enron and Lucene) and use query frequencies grabbed from Google Trends. We give the adversary imperfect auxiliary information to run the attack.

Enron: 30k emails

Lucene: 66k emails

From Google Trends



freq



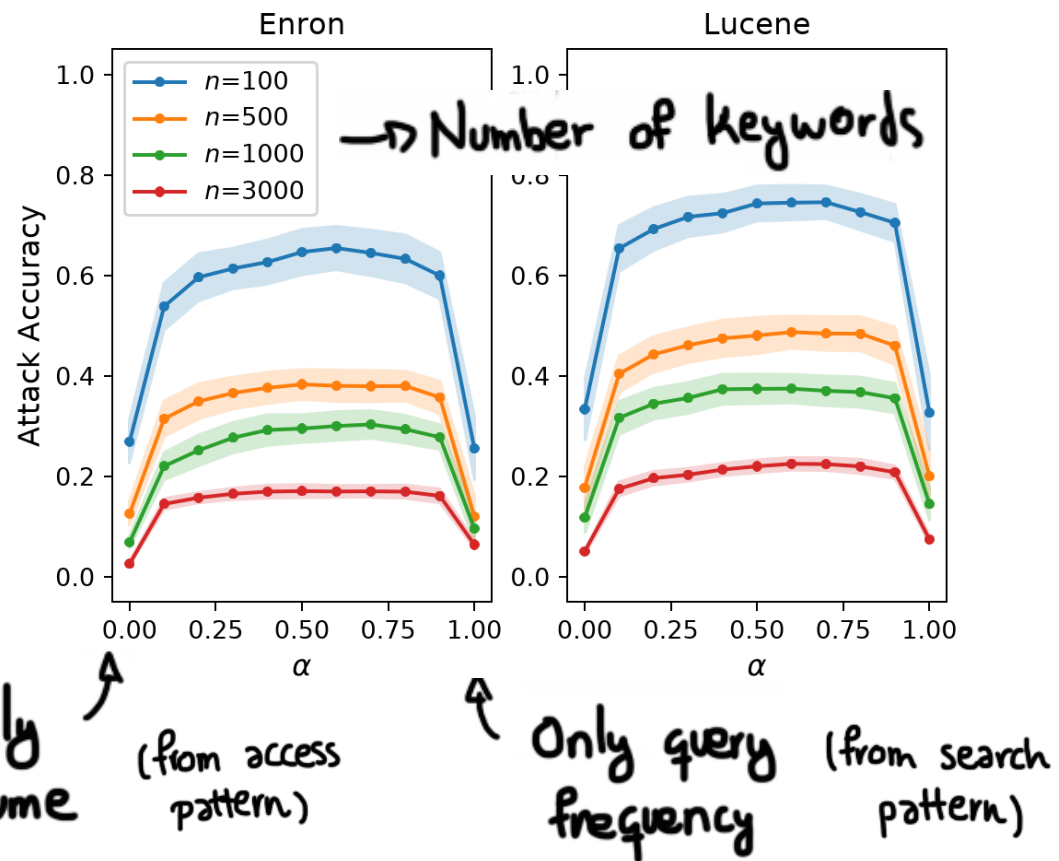
Aux info has 5-week offset

Results I

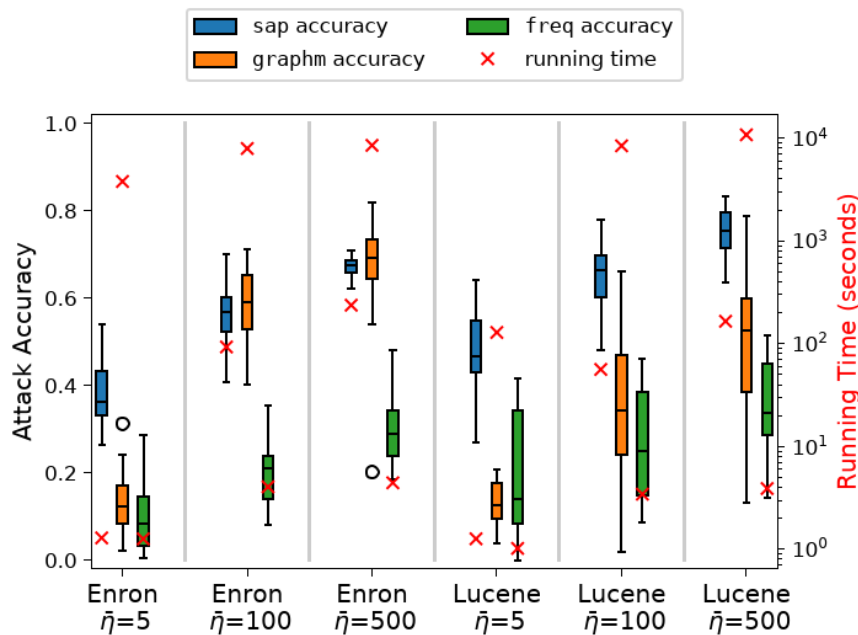
5 queries / week
50 weeks

Importance of search+access pattern

By combining both volume and frequency information, SAP achieves high query recovery (left).
SAP outperforms current state-of-the-art attacks (right).



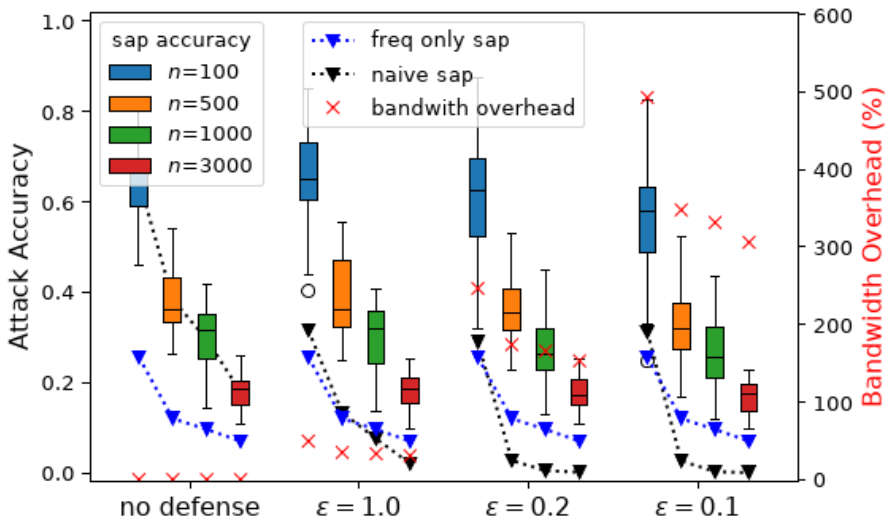
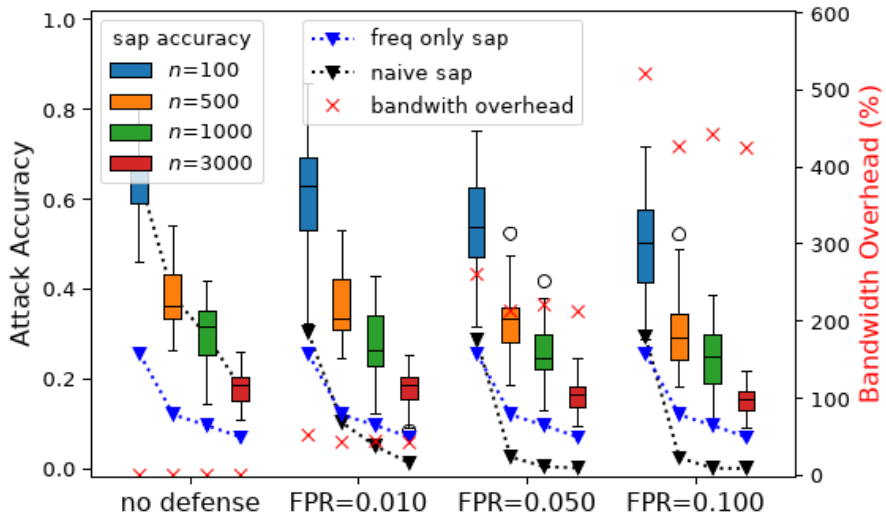
Attack comparison



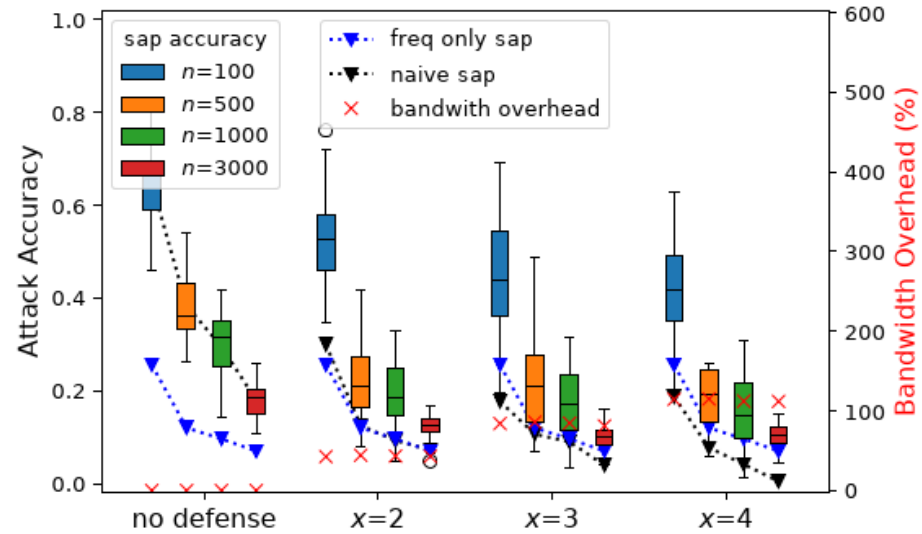
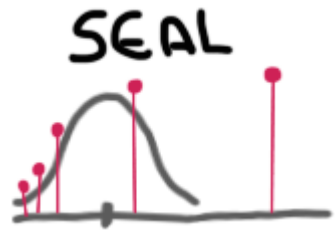
freq: Liu et al. "Search pattern leakage in searchable encryption: Attacks and new construction". Information Sciences, 2014.

graphm: Pouliot and Wright. "The shadow nemesis: Inference attacks on efficiently deployable, efficiently searchable encryption". CCS'16

Results II



By adapting SAP against previous defenses, we are able to practically bypass two of them (CLRZ and PPYY) and we still achieve non-trivial recovery rates for the third one (SEAL).



Conclusions

- ▷ **SAP** is efficient and strong
- ▷ Frequency + Volume leakage is dangerous
- ▷ Padding strategies are not very effective
- ▷ Hiding search pattern and/or frequency is important



Our results show the importance of hiding search patterns and/or frequency leakage. Recent works that are moving in this direction seem promising.

There is hope!

PANCAKE
Grubbs et al.
USENIX'20

OSSE
Shang et al.
NDS5'21

SW:SSSE
Gui et al.



Thank you!



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