



Weakly Randomized Encryption

And the Strength of Weak Randomization

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“Executive” Summary

Weakly Randomized Encryption

- A safer upgrade to deterministic encryption
- Secure against most common “snapshot” attacks
- Easy to deploy
- ACID properties*
- Low overhead



Research Questions

1. What security can we achieve if **easy deployability** is a **hard constraint**?
2. Are there PPE-like constructions that provide **any meaningful security** against inference???

RELATED WORK

Property-Preserving Encryption (PPE)

- Deterministic and Efficiently Searchable Encryption [BBO07,ABO07]
- CryptDB [PRZB11]
- Microsoft SQL Server “Always Encrypted”


Parallel Invention

- [LP18] Lacharité and Paterson. Frequency Smoothing Encryption: Preventing snapshot attacks on deterministically encrypted data.
 - <https://eprint.iacr.org/2017/1068>
 - Most similar to our *Proportional Salt Allocation*


Inference Attacks

1. Offline inference (the “snapshot” model)
 - IKK12, NKW15
 - CGPR15, GSBNR17
2. Online inference
 - KKNO16, LMP18
 - GLMP18, GLMP19
3. Inference from database/OS artifacts
 - GRS17


Defense Against Inference Attacks

1. Offline inference:
 - IKK12, NKW15
 - CGPR15, GSBNR17

Focus of this work

 - Defend against the **most common attacks** (i.e. snapshots / SQL injection)
 - Maximize backwards compatibility
 - What security & performance can we get?
2. Online inference
 - KKNO16, LMP18
 - GLMP18, GLMP19

Harder problem / Future work

 - Attacks apply to stronger constructions too
3. Inference from database/OS artifacts
 - GRS17

Mostly engineering??

 - Not worth trying to fix this if you can't also defend #1

SECURITY GOALS

Security Game



$$D_0 = (m_{0,0}, m_{0,1}, \dots, m_{0,n})$$

$$D_1 = (m_{1,0}, m_{1,1}, \dots, m_{1,n})$$



$$b = \{0,1\}^1$$

$$EDB = \text{Enc}(\text{Shuffle}(D_b))$$



$$b'$$



Adversary wins iff $b' == b$

Statistical Distance and Security

Definition 3 (Statistical Distance). *The statistical distance Δ between two random variables X, Y over a common domain ω is defined as:*

$$\Delta(X, Y) = \frac{1}{2} \sum_{\alpha \in \omega} \left| \Pr(X = \alpha) - \Pr(Y = \alpha) \right|$$

Definition 4 (Distinguishing Two Distributions). *Let P_0 and P_1 be probability distributions on a finite set R . Then for every adversary \mathcal{A} , we have the distinguishing advantage of \mathcal{A} between P_0 and P_1 ,*

$$\Pr[\text{Dist}_{\mathcal{A}}(P_0, P_1)] \leq \Delta(P_0, P_1)$$

CONSTRUCTIONS

Efficiently Searchable Encryption

[BBO07, ABO07]

Plain Table

Row ID	Animal
1	Dog
2	Horse
3	Cat
4	Cat
5	Dog
6	Horse
7	Dog
8	Dog
9	Cat

Efficiently Searchable Encryption

[BBO07, ABO07]

Plain Table

Row ID	Animal
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Encrypted Table

Row ID	Tag	Cipher
1	F(Dog)	E(Dog)
2	F(Horse)	E(Horse)
3	F(Cat)	E(Cat)
4	F(Cat)	E(Cat)
5	F(Dog)	E(Dog)
6	F(Horse)	E(Horse)
7	F(Dog)	E(Dog)
8	F(Dog)	E(Dog)
9	F(Cat)	E(Cat)

Efficiently Searchable Encryption

[BBO07, ABO07]

Plain Table

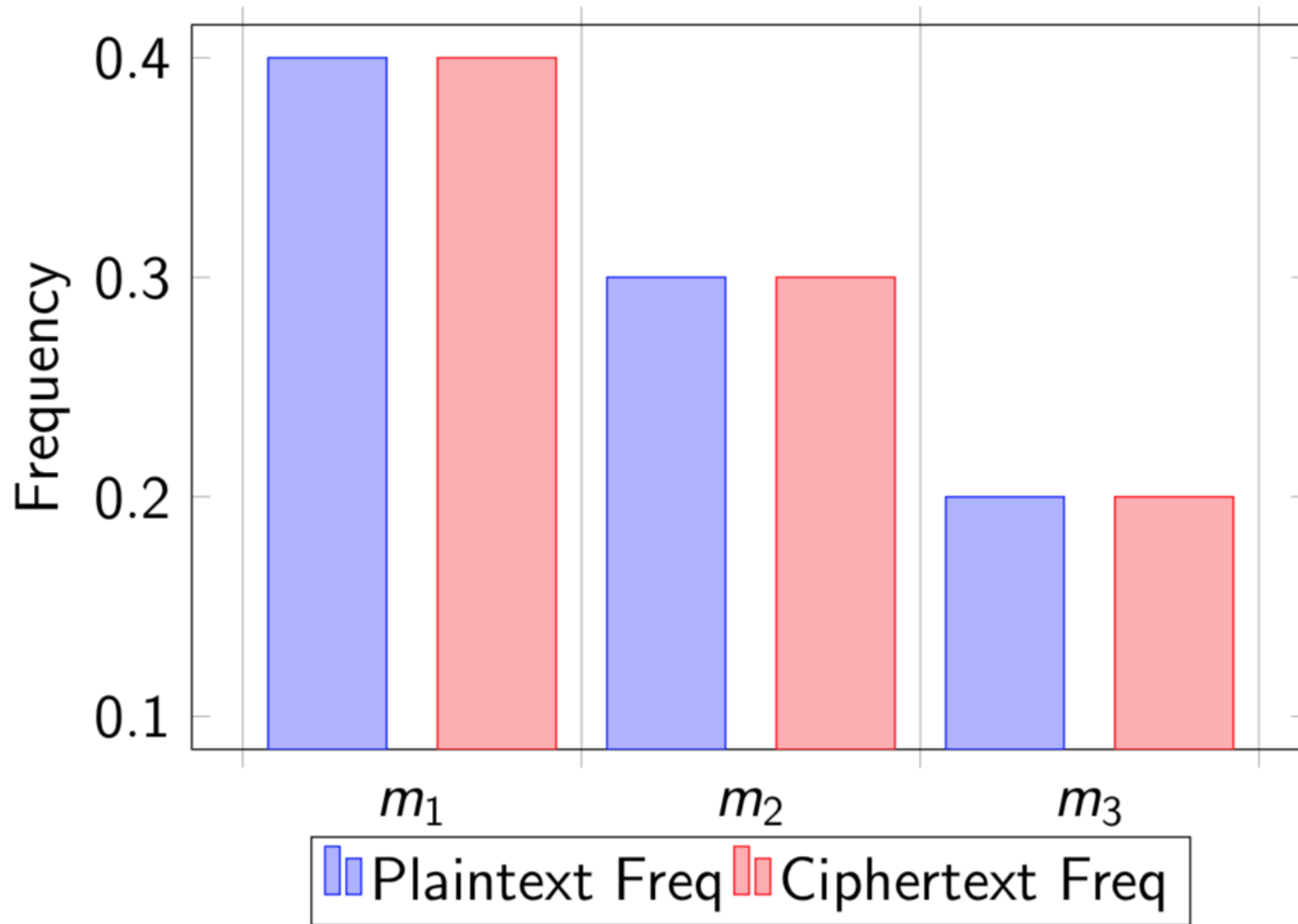
Row ID	Animal
1	Dog
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Encrypted Table

Row ID	Tag	Cipher
1	eb3f	653c
2	137a	bb21
3	6f20	e0f3
4	6f20	9201
5	eb3f	bbcf
6	137a	d830
7	eb3f	c971
8	eb3f	ee26
9	6f20	7a0b

Deterministic Ciphertext Frequencies



Randomizing Deterministic Encryption



- Too random → Not useful ☹️



- Too predictable → Not secure ☹️



- Just enough randomness → 😊

To Encrypt

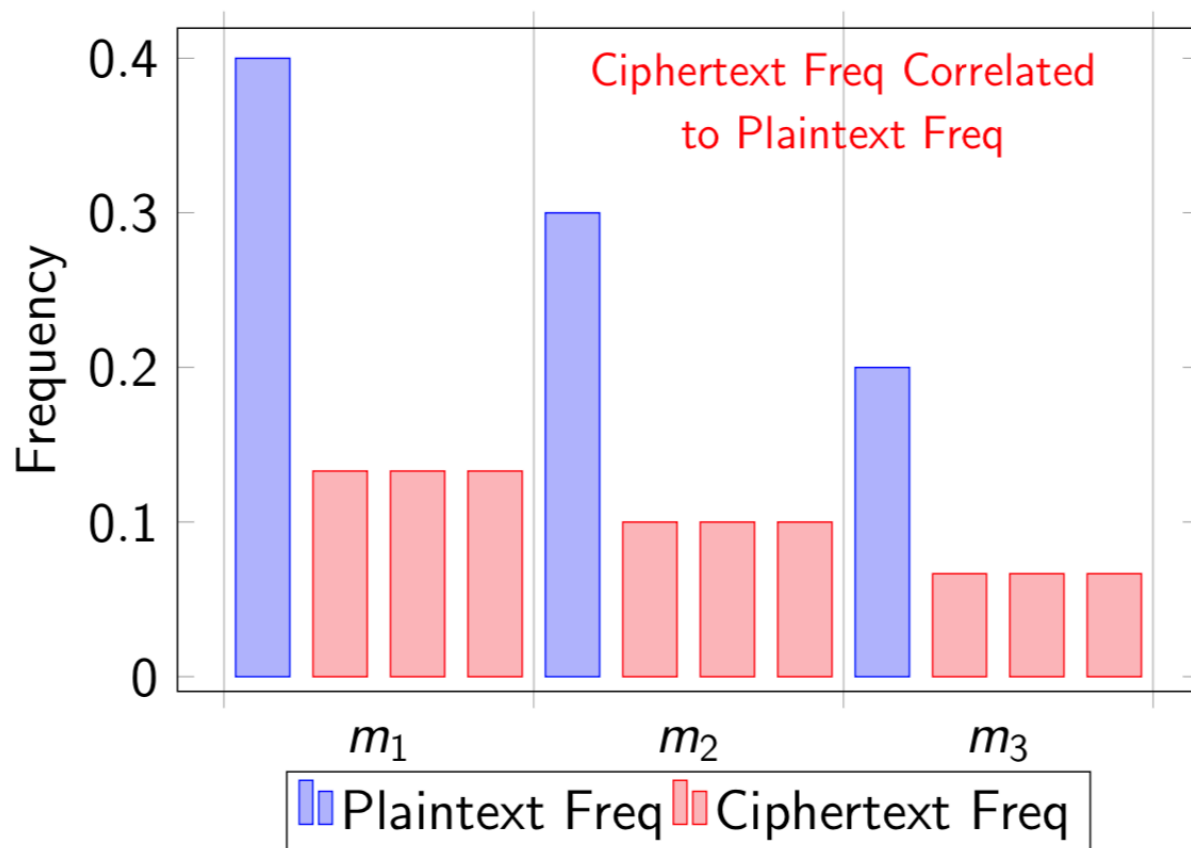
1. Choose random, **low entropy** salt s
2. Tag $t = F_{k1}(s || m)$
3. (Randomized) ciphertext $c = E_{k2}(m)$

To Search

1. Generate all possible tags for msg m
 - For each salt s_i :
Let $t_i = F_{k1}(s_i || m)$
2. Encrypt query
 - SELECT ...
FROM enc_table
WHERE tag in (t_1, t_2, \dots, t_n);

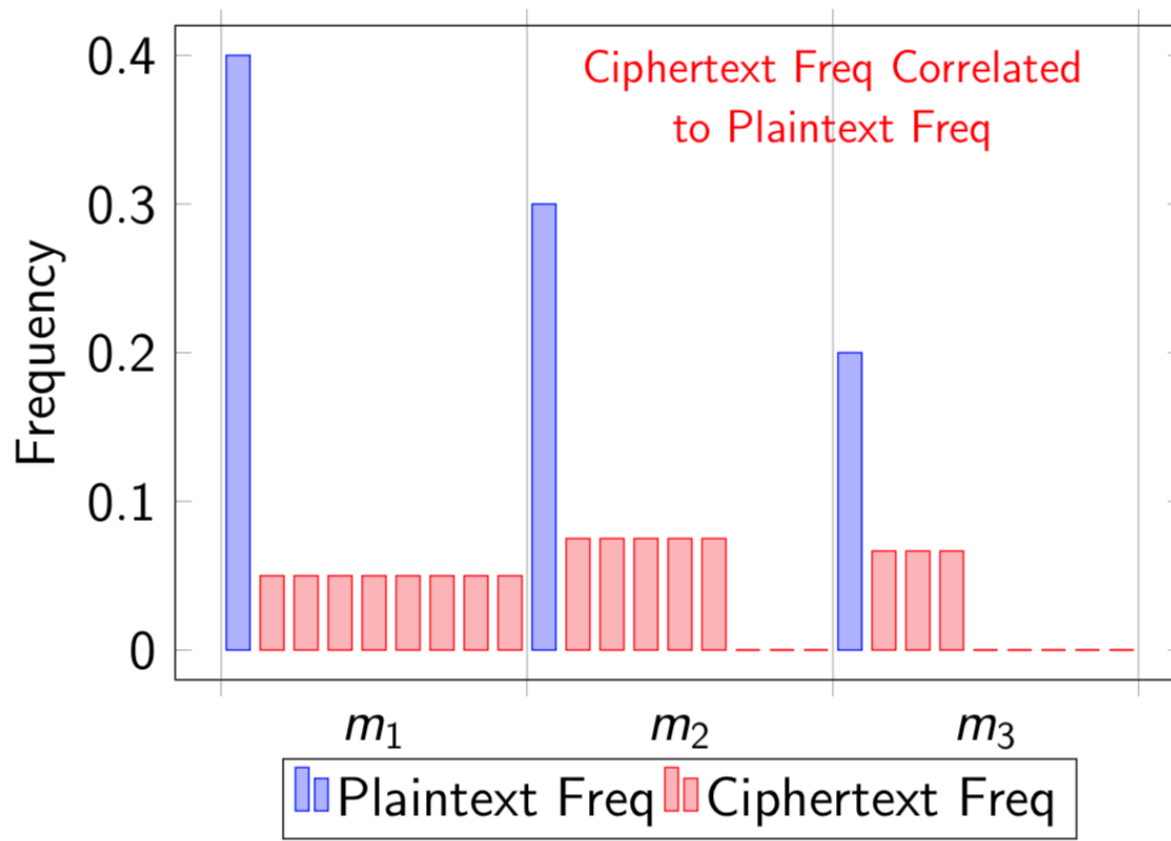
Strawman Construction: Fixed Salts

- Choose salt uniformly from $[1..N]$
 - e.g. $N = 3$



Proportional Salt Allocation

- Allocate salts in proportion to frequency



Frequencies are closer to Uniform

Some *aliasing* effects

Poisson Salt Allocation

Question:

How to allocate message m 's probability mass to the ciphertexts?

0

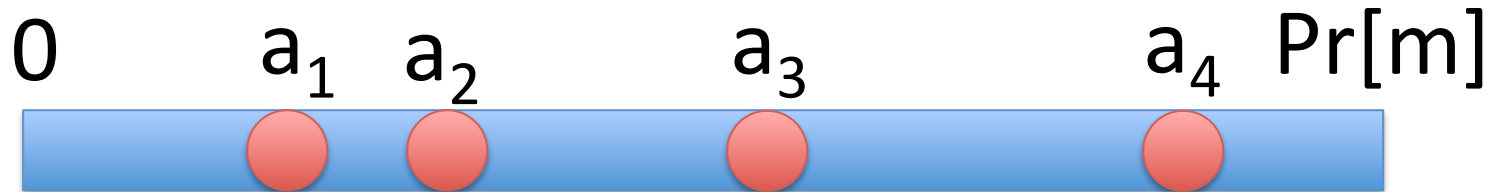
$\text{Pr}[m]$



Poisson Salt Allocation

Idea:

Sample points from a Poisson process w rate param λ

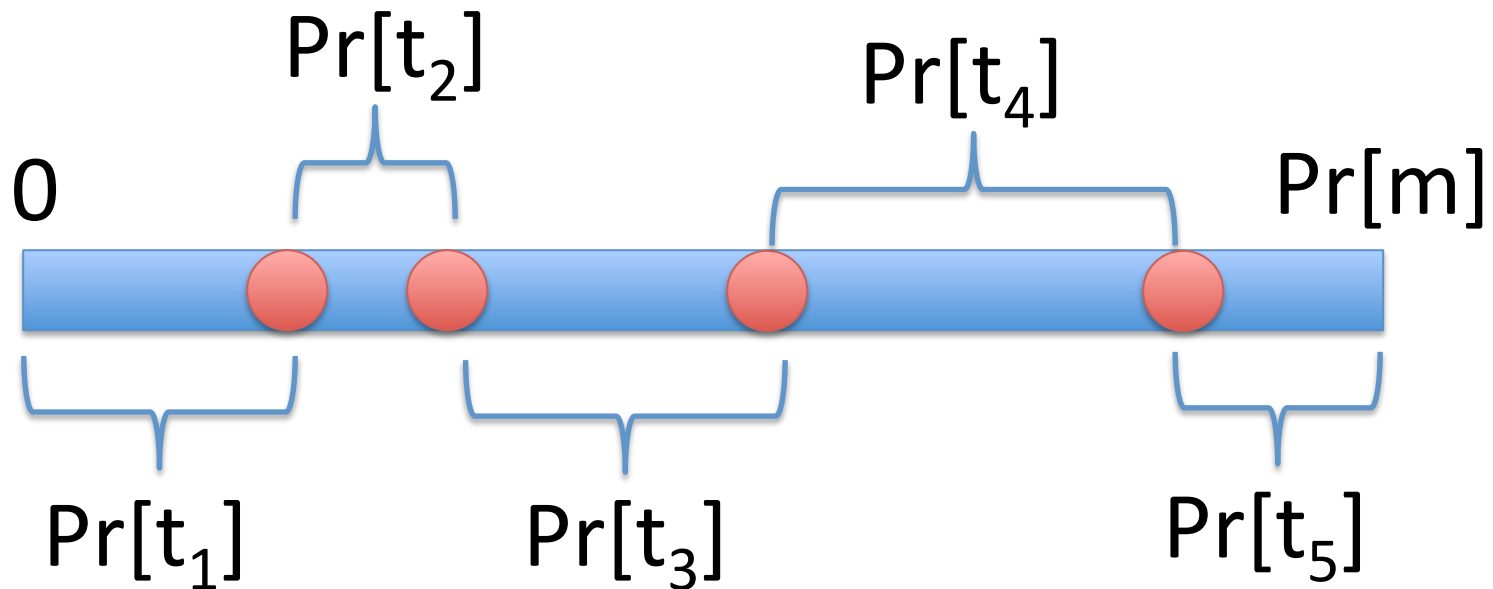


Poisson Salt Allocation

Idea:

Sample points from a Poisson process w rate param λ

Distances between points (“inter-arrivals”) give tag frequencies



Poisson Security

- Ciphertext freqs are **identically distributed!**
 - $\Pr[t_j] \sim \text{Exponential}(\lambda)$ for all j

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- No statistical distance \rightarrow No guessing advantage

Poisson Security

- Ciphertext freqs are **identically distributed!**
 - $\Pr[t_j] \sim \text{Exponential}(\lambda)$ for all j

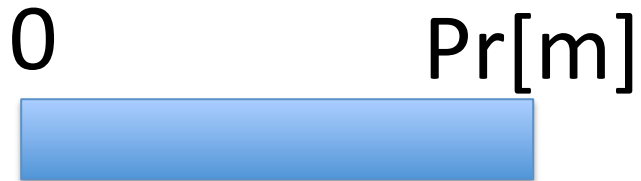
Whoops... Not quite true..

- Identical distribution *They are **almost identically distributed.** :-*
- No statistical distance \rightarrow No guessing advantage

Something Fishy About Poisson

Problem:

What if there are no arrivals in the interval $[0, \text{Pr}[m]]$???

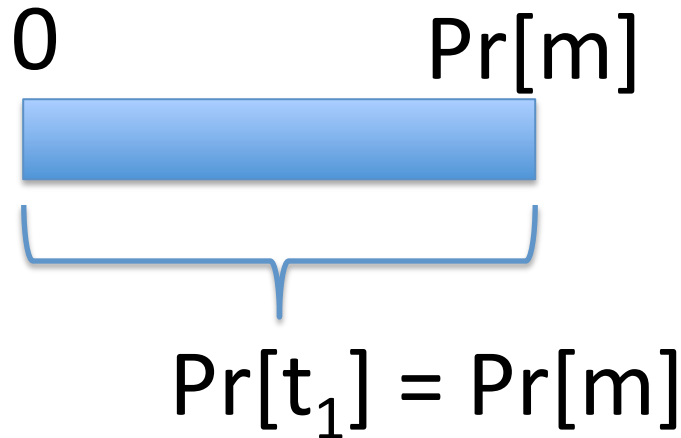


Something Fishy About Poisson

Problem:

What if there are no arrivals in the interval $[0, \text{Pr}[m]]$???

No choice but to give all of m 's probability mass to a single tag



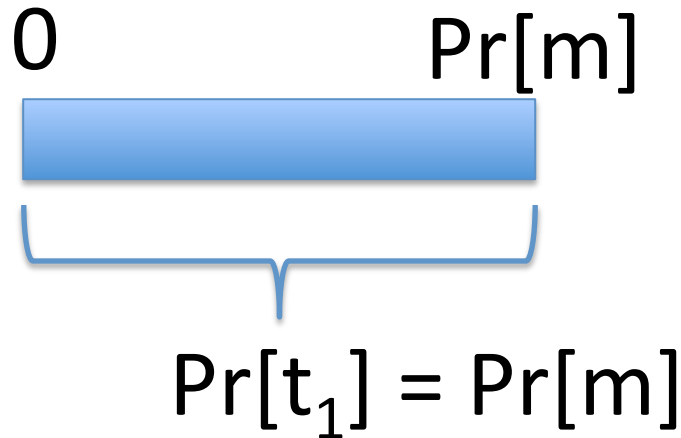
Something Fishy About Poisson

Problem:

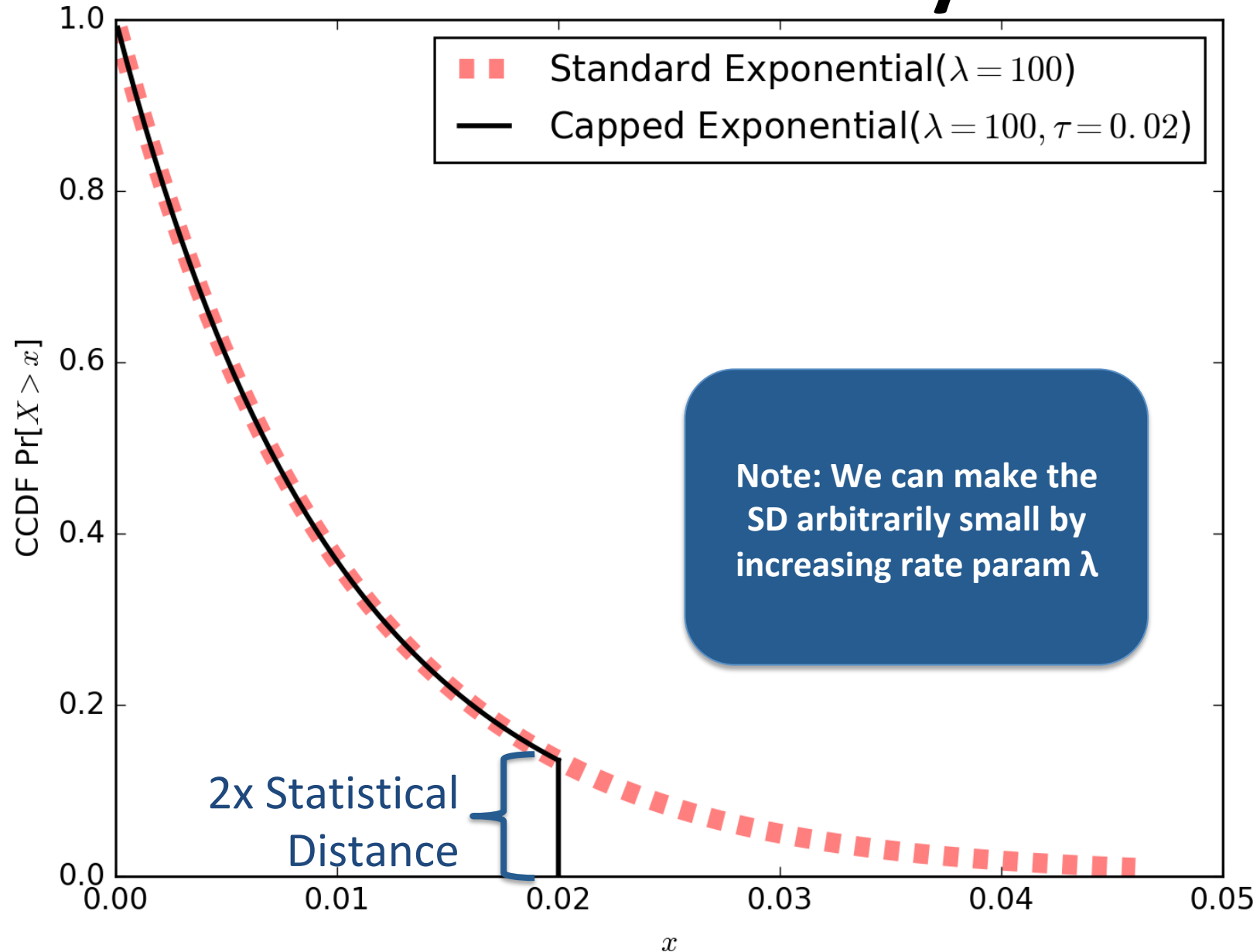
What if there are no arrivals in the interval $[0, \text{Pr}[m]]$???

No choice but to give all of m 's probability mass to a single tag

Not really a true Exponential. Can the Adv now distinguish?



Poisson: Security

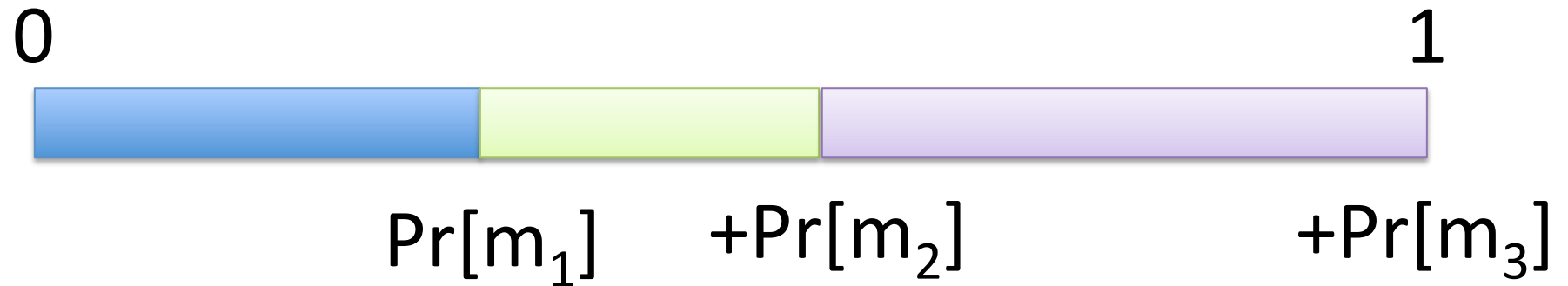


Poisson: One More Problem

- Lacharite-Paterson attack: What if Adv looks at more than one ciphertext?
 - Goal: Find a set of search tags t_1, t_2, \dots, t_n s.t.
 - $\Pr[m] = \sum_j \Pr[t_j]$
 - These records are *probably* (???) the encryptions of m
 - Difficulty: Bin packing problem :-\
 - On the bright side:
 - Might be a hard (NP) instance
 - Solution might (tend to) select the wrong records

Bucketized Poisson

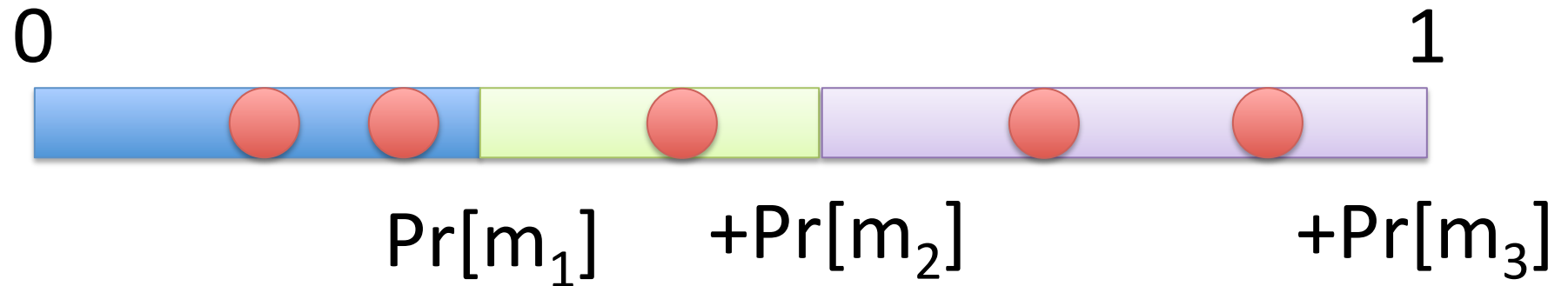
Lay out plaintext freqs on the number line $[0..1]$



Bucketized Poisson

Lay out plaintext freqs on the number line $[0..1]$

Sample from the Poisson process

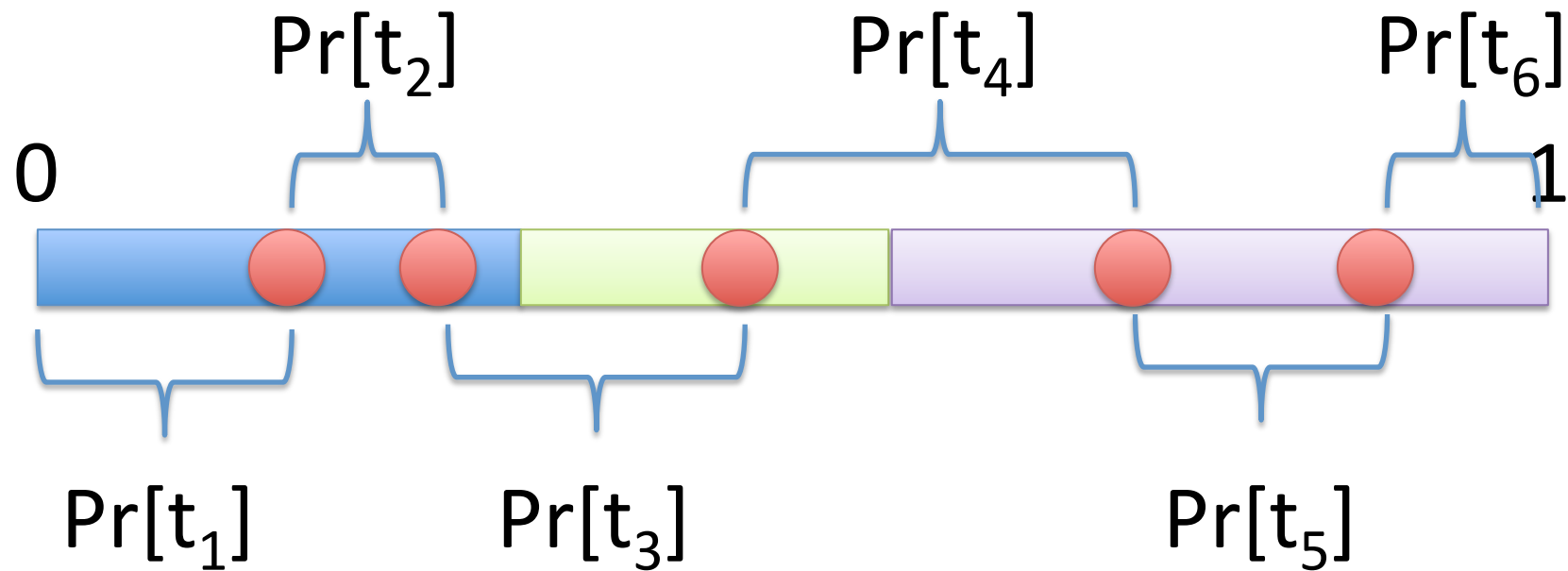


Bucketized Poisson

Lay out plaintext freqs on the number line $[0..1]$

Sample from the Poisson process

Use inter-arrivals to fix a set of search tags for **all** plaintexts to share

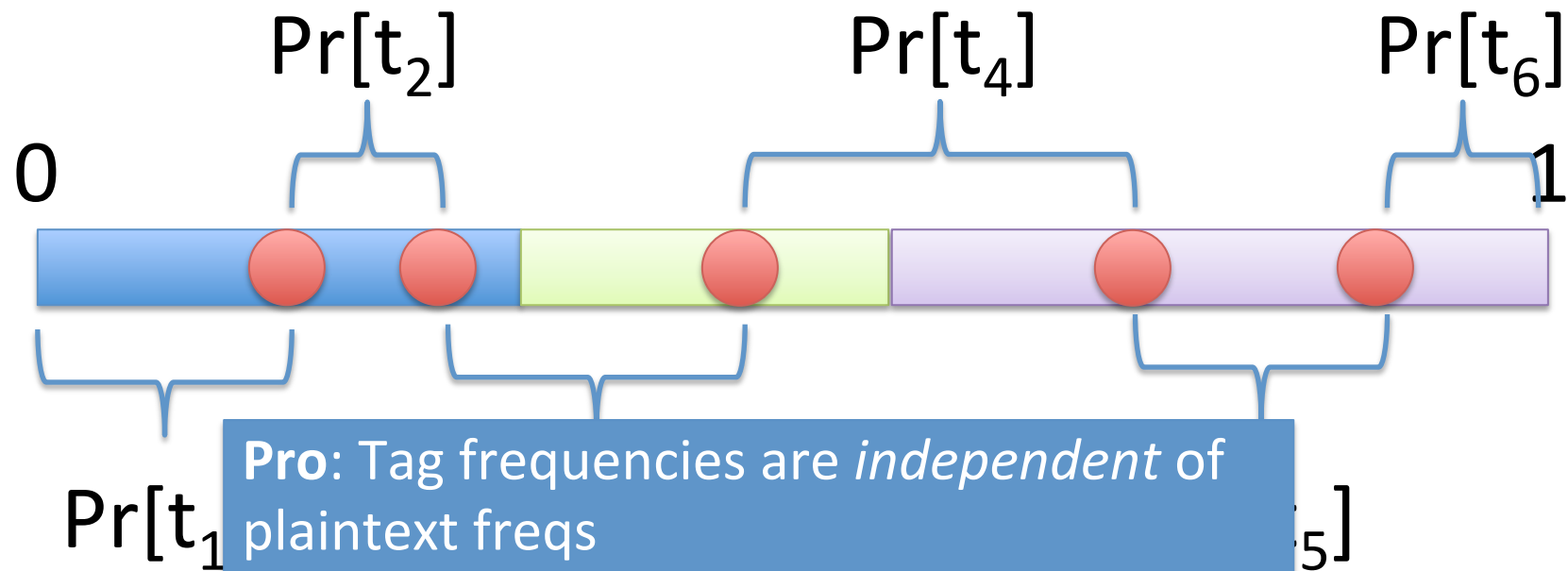


Bucketized Poisson

Lay out plaintext freqs on the number line $[0..1]$

Sample from the Poisson process

Use inter-arrivals to fix a set of search tags for **all** plaintexts to share



Pro: Tag frequencies are *independent* of plaintext freqs

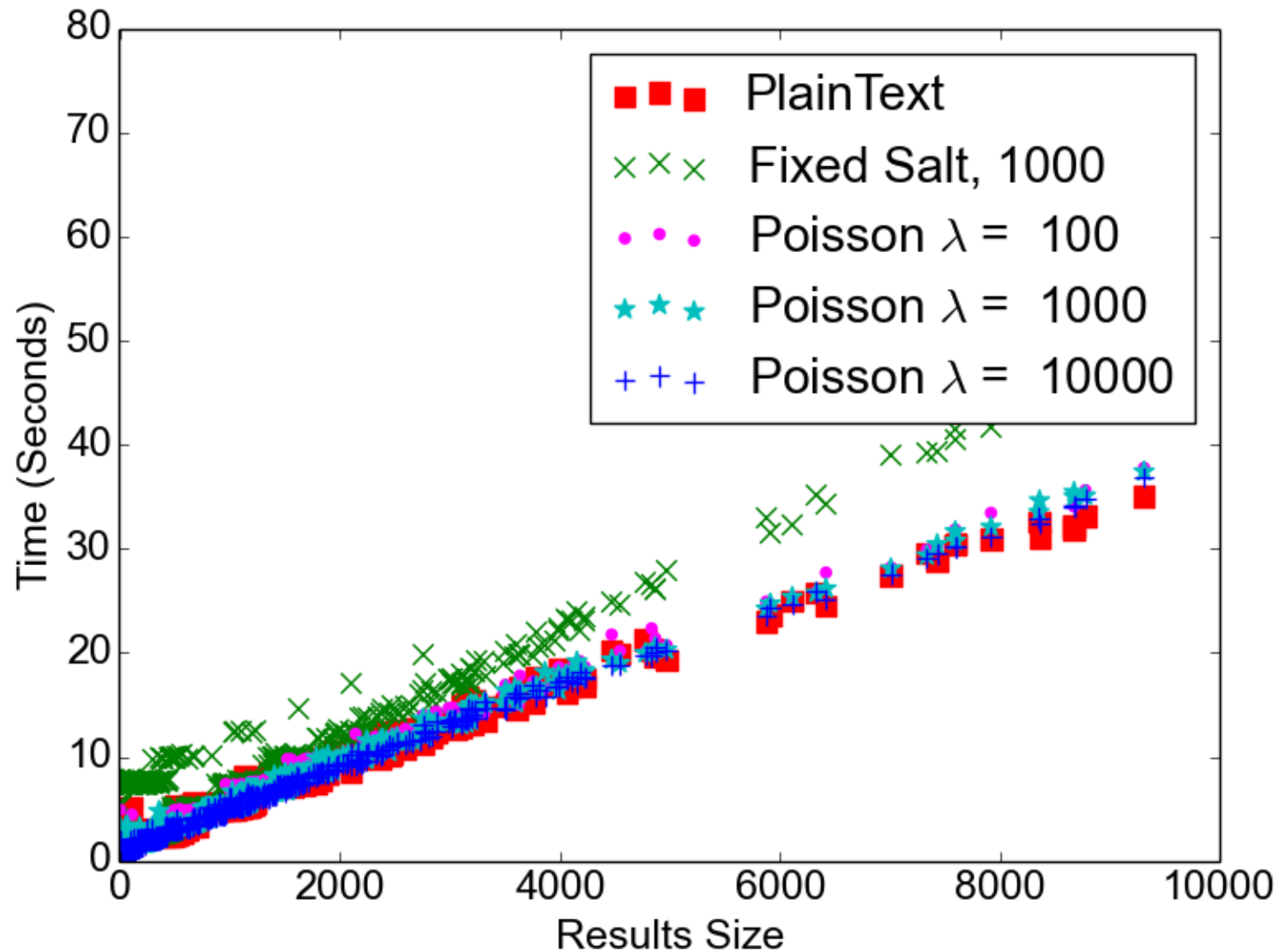
Con: Tags are now *buckets* representing multiple plaintexts

EMPIRICAL EVALUATION

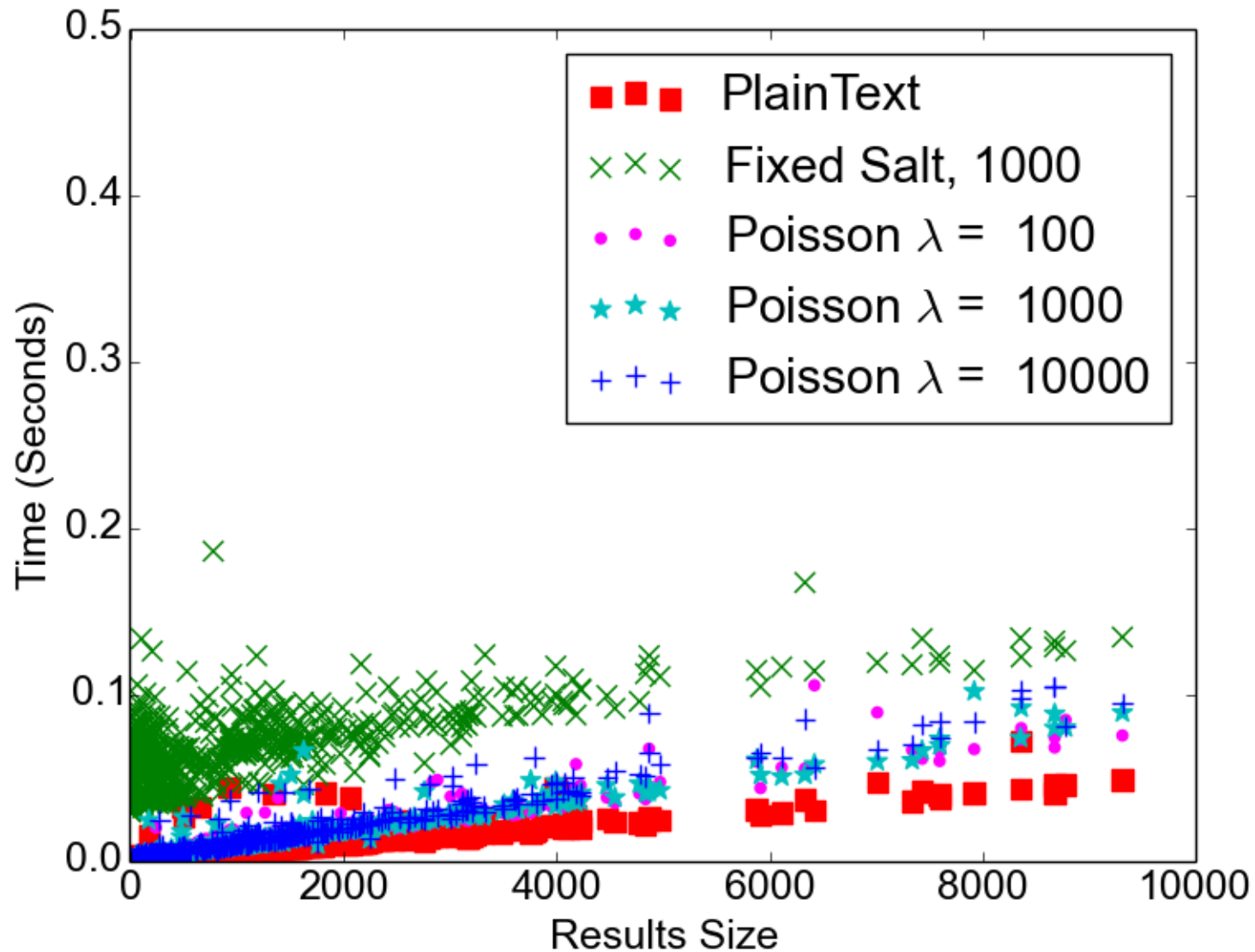
Experimental Procedure

- Used SPARTA testing framework from MIT-LL
 - Generated synthetic databases
 - 1M, 10M records
 - Generated synthetic queries
 - `SELECT ... FROM table WHERE column = value;`
 - Return up to 10k matching records
- Ran queries on real SQL databases
 - Google Compute Engine
 - Local Postgres server

Performance: Cold Cache



Performance: Warm Cache



Conclusion

- WRE Contributions
 - Easy to deploy
 - Secure against most common threats
 - Performance close to plaintext
- Future Work / Open Problems
 - Security for queries? For access pattern?
 - Security for multiple (correlated) columns?
 - Range queries?