DBREACH

Database Reconnaissance and Exfiltration via Adaptive Compression Heuristics

Mathew Hogan
Stanford University

Saba Eskandarian
UNC Chapel Hill

Yan Michalevsky
Anjuna Security

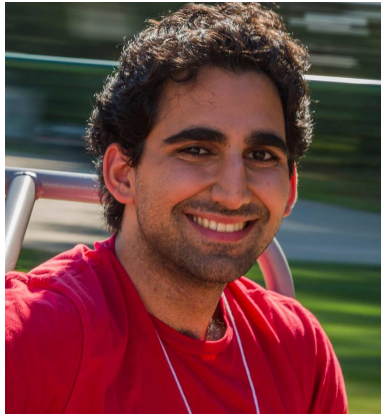
Who We Are

Mathew Hogan

- MS Candidate in CS at Stanford, Security track
- BS in CS from Stanford, Systems track

Yan Michalevsky

- CTO and co-founder at Anjuna.io
- PhD in Security and Crypto from Stanford



Saba Eskandarian

- Assistant Prof. at UNC Chapel Hill
- PhD in Crypto and Security from Stanford

Outline

1. Background
2. Our Attack
3. Roadblocks & Optimizations
4. Analysis
5. Mitigations
6. Conclusion

Background

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Key idea: use compression to reveal information about the original content

John Kelsey. "Compression and Information Leakage of Plaintext," FSE 2002.

CRIME/BREACH (2012/13)

Secret included in encrypted and compressed messages between client and server

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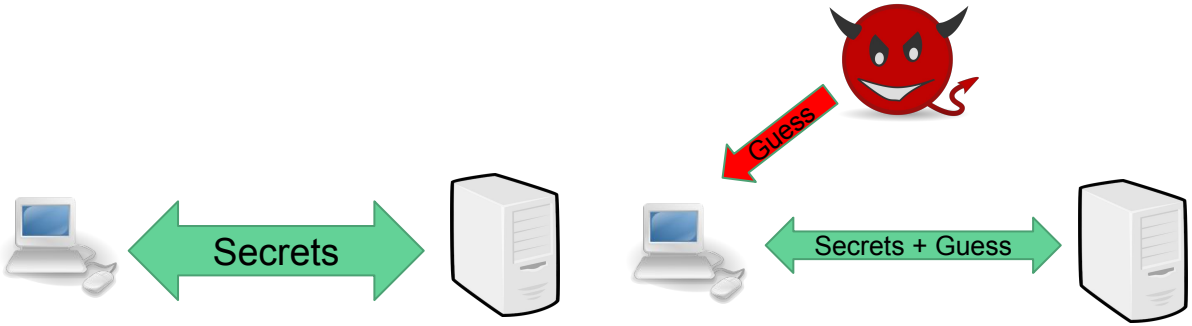

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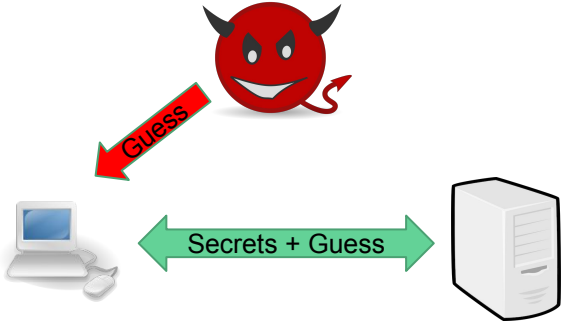

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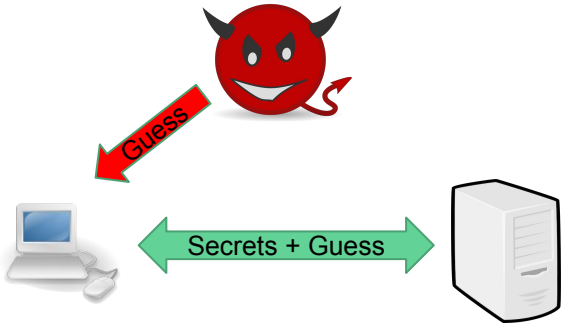

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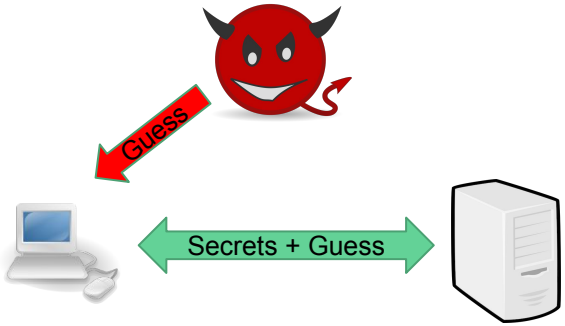

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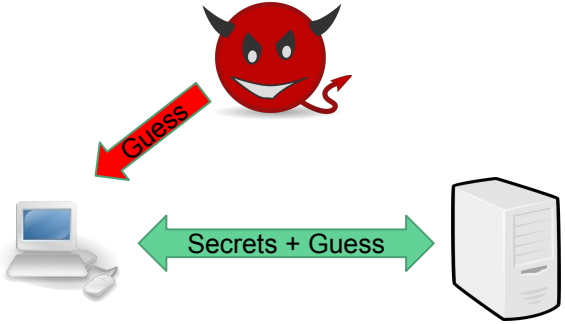

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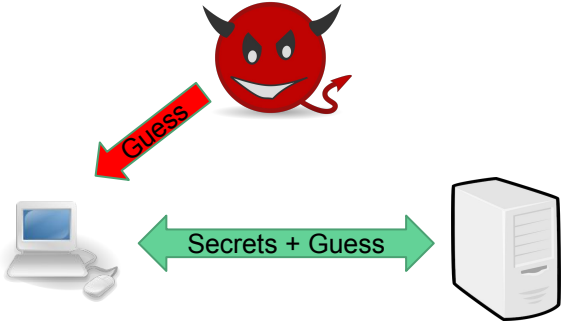

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Where else do all these factors come together?

DBREACH in a Nutshell

Compression side-channel attack against databases

Attacker recovers other users’ encrypted content

Extends techniques from CRIME/BREACH beyond TLS to database context

MariaDB/InnoDB Encryption and Compression

Data-at-Rest Encryption - transparently encrypt data before writing to disk

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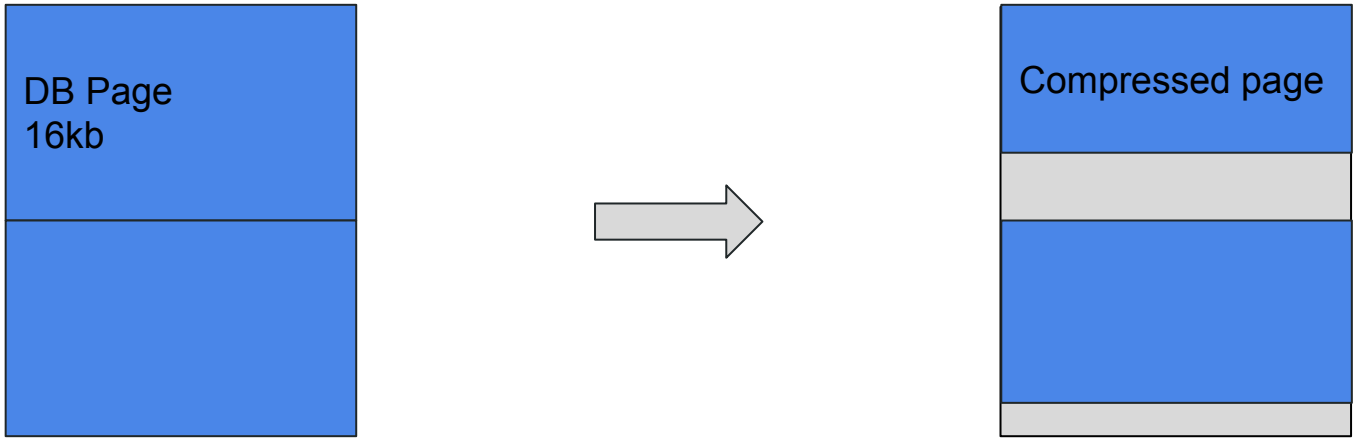

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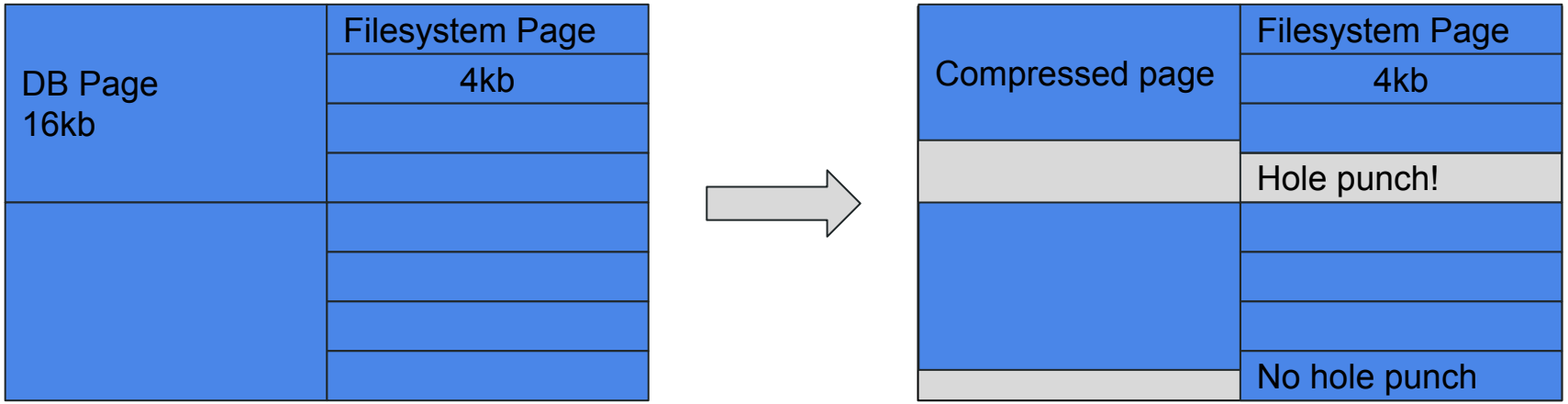

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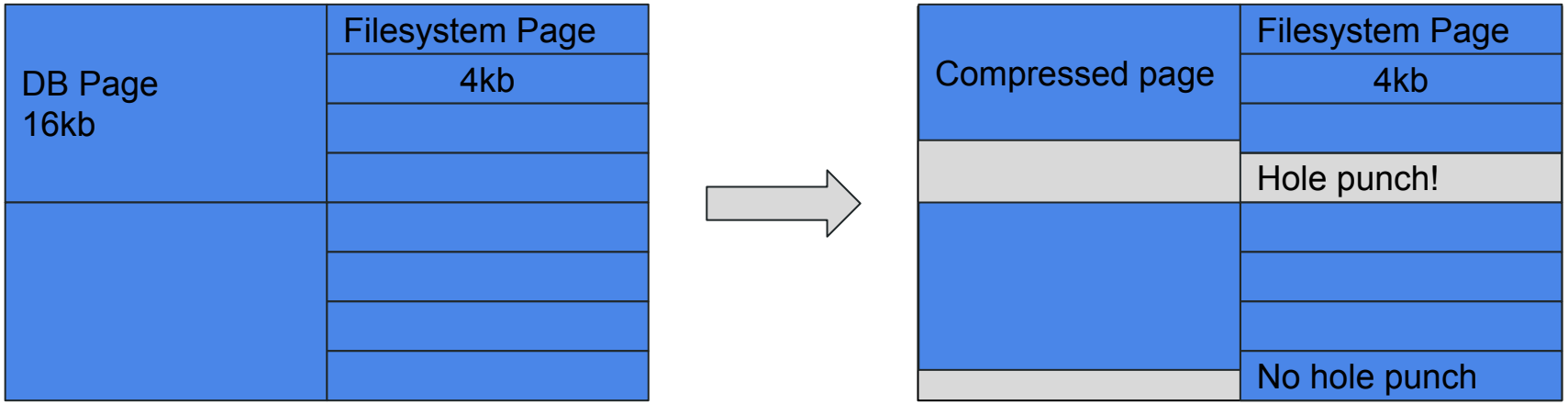
Uses *hole punching* to save space, only helps when there is enough compression to remove a whole filesystem page



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  zlib additionally has a Huffman Coding step

Our Attack

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If UPDATE permissions can’t be achieved, an attacker with write access can force an update by rolling back the table file and inserting.

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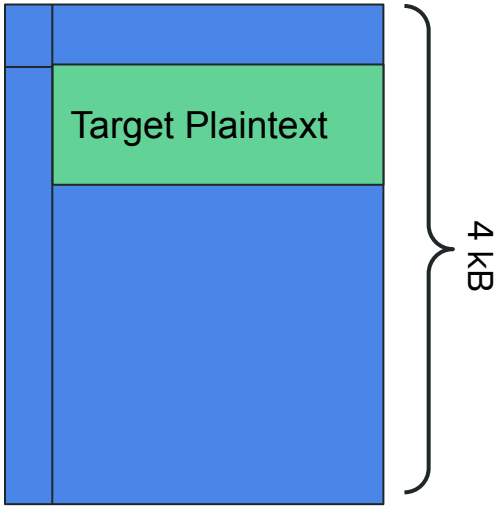

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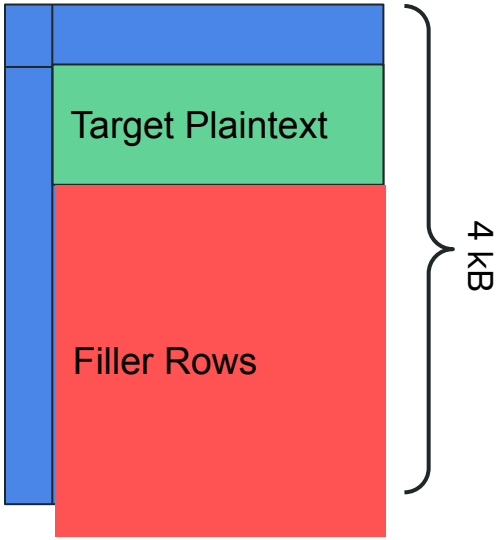

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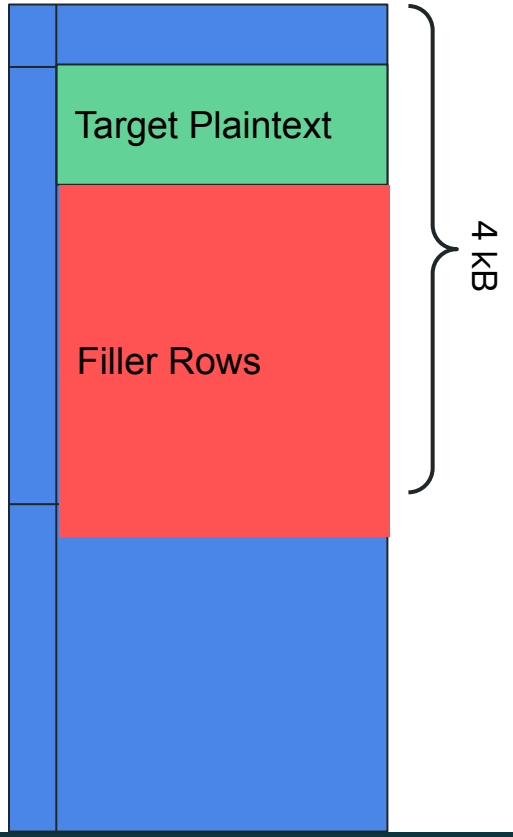

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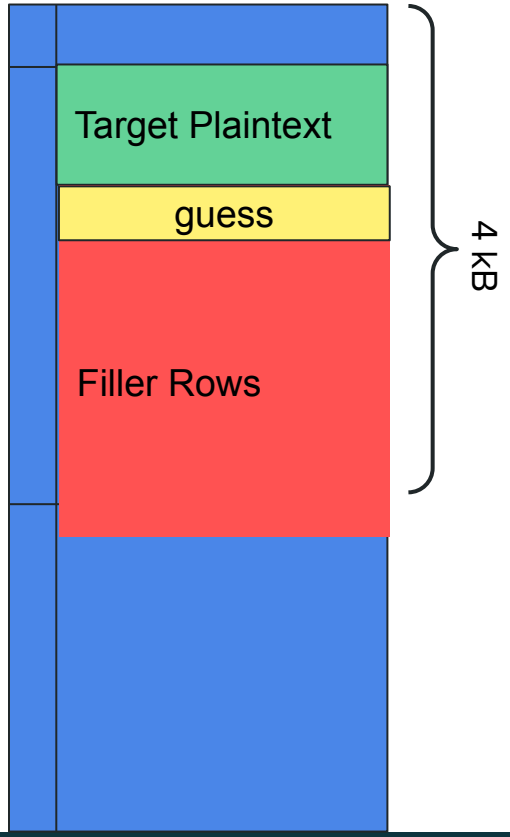

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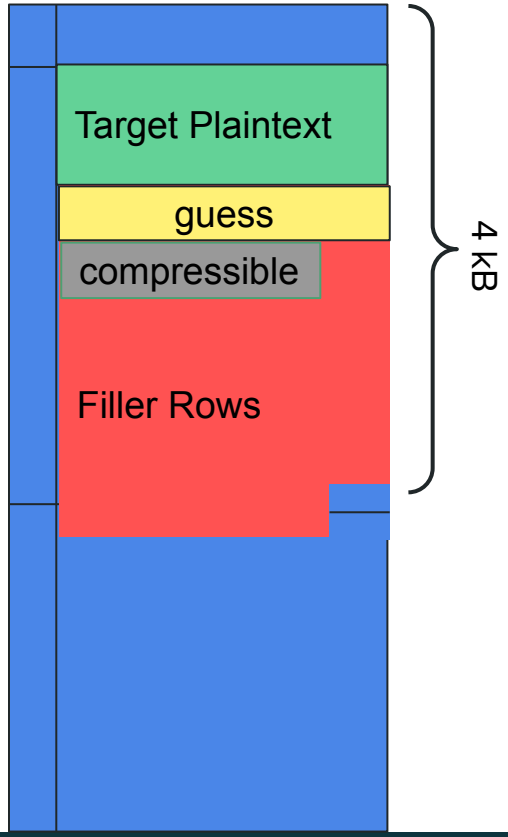

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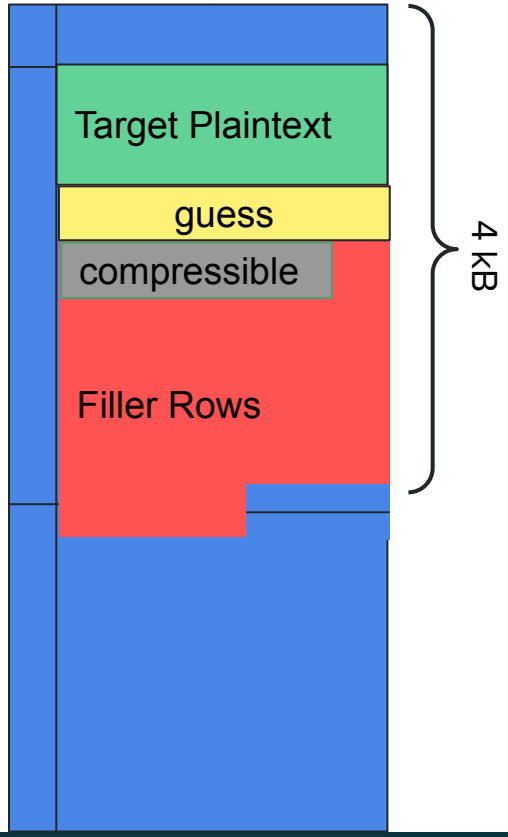

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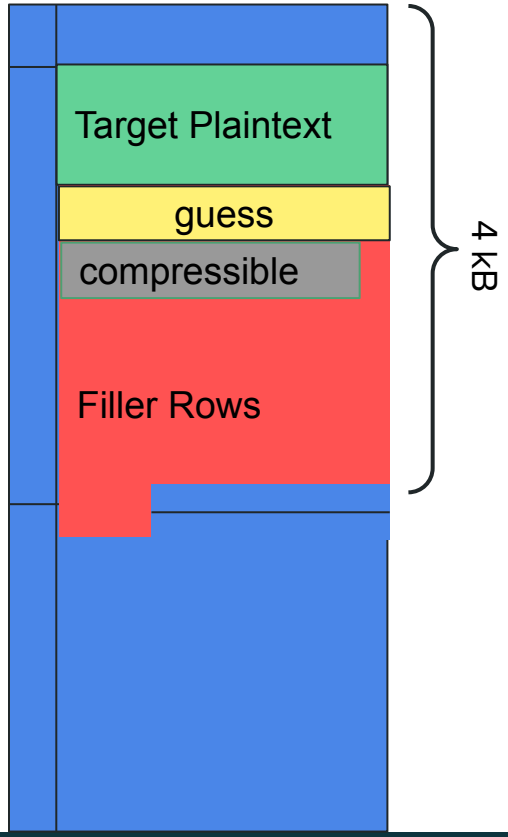

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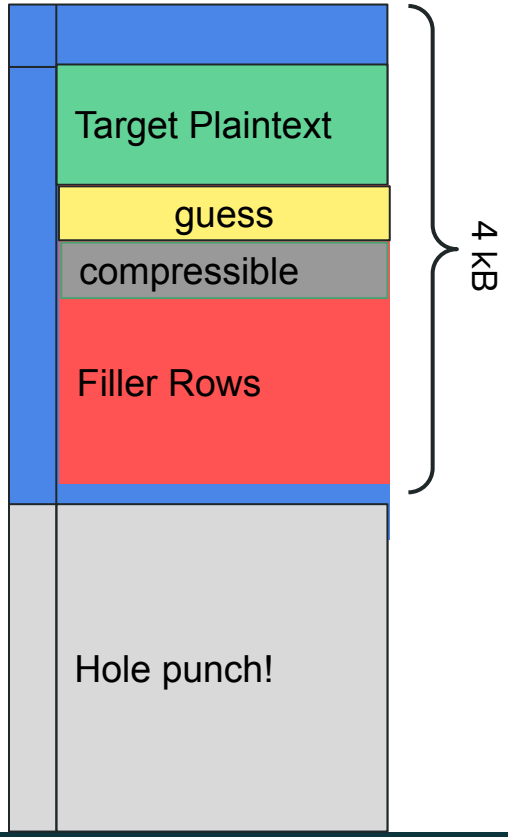

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3. Byte by byte, make filler rows compressible
    a. The number of bytes until the table shrinks determines this guess’s “compressibility score”

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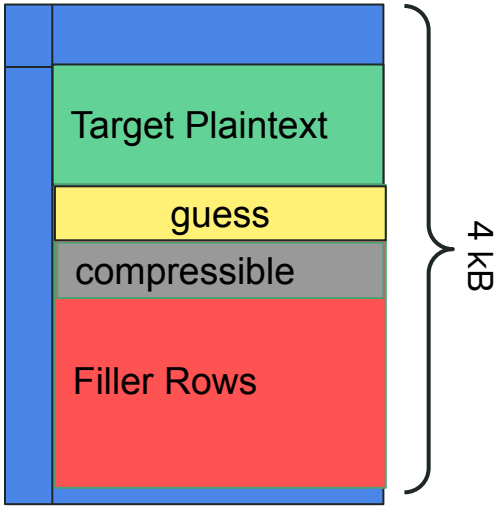

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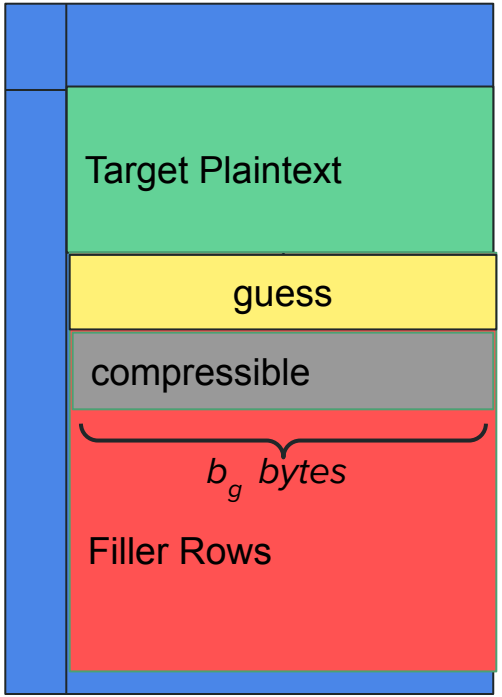

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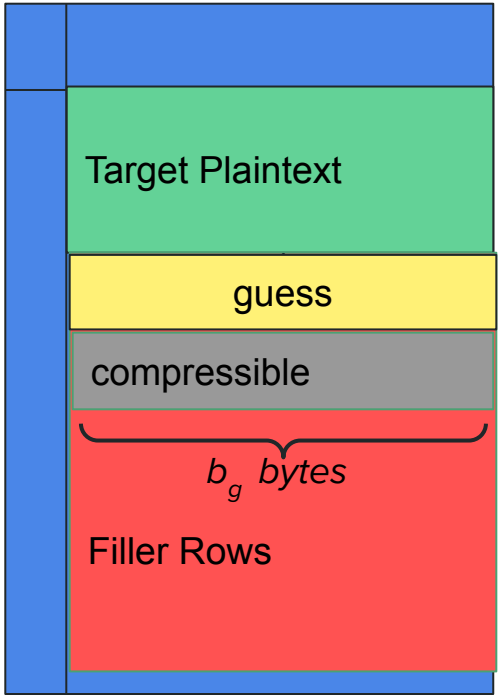
The compressibility score c_g is calculated as follows:

$$c_g = 1 / b_g$$



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  - More complicated
  - We need a reference point for compressibility scores

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To gain reference points, we calculate scores s_{yes} and s_{no} for strings we know to be in and not in the table, respectively.

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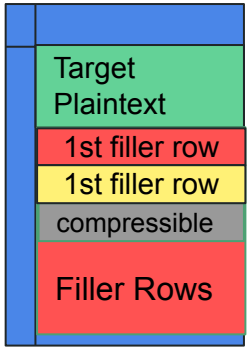

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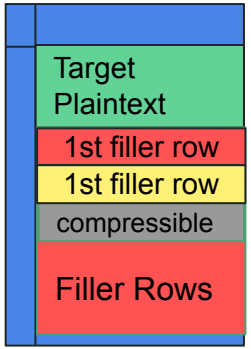

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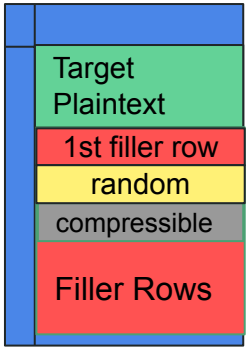

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If a guess's score is within some threshold of s_{yes}, answer "yes"

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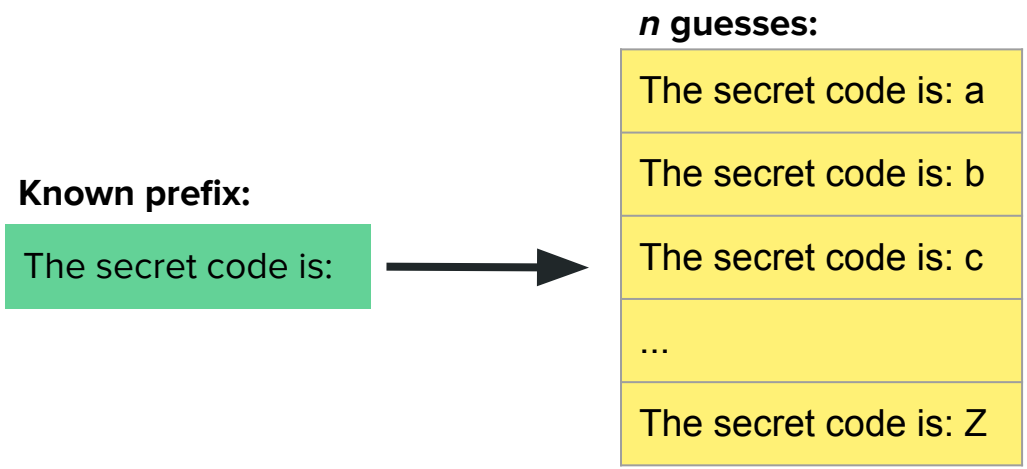

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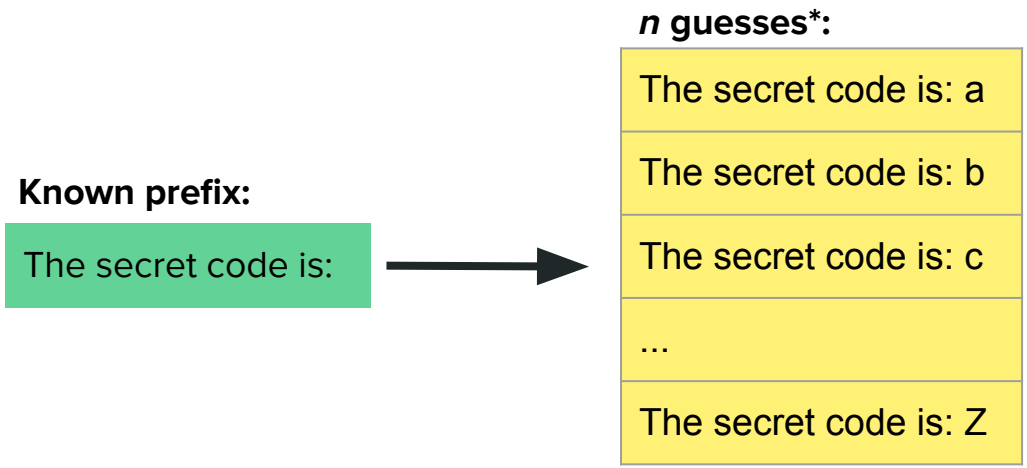

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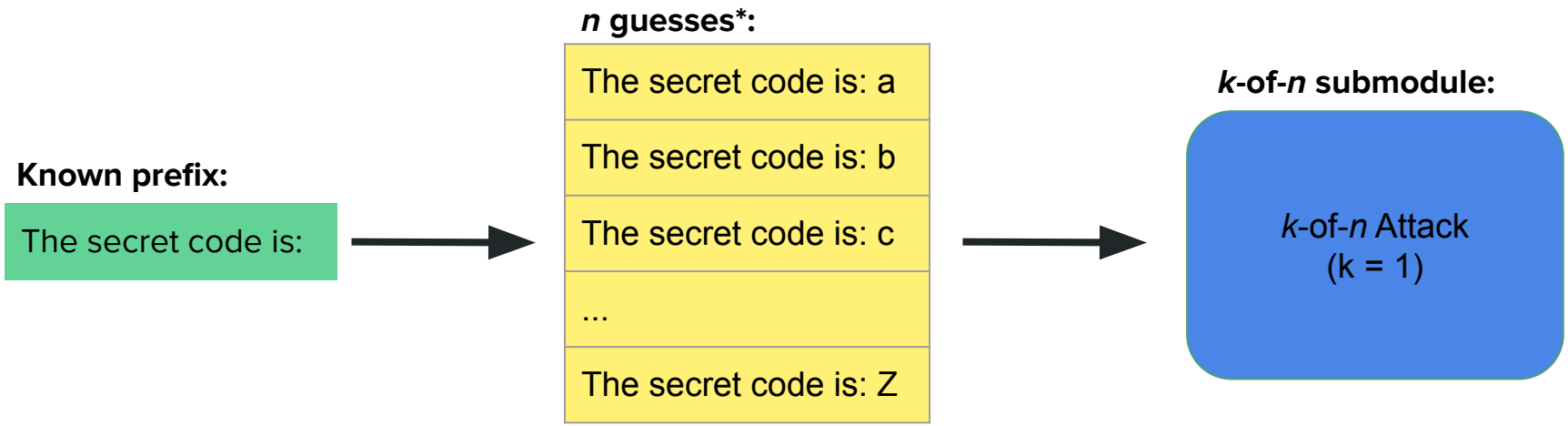

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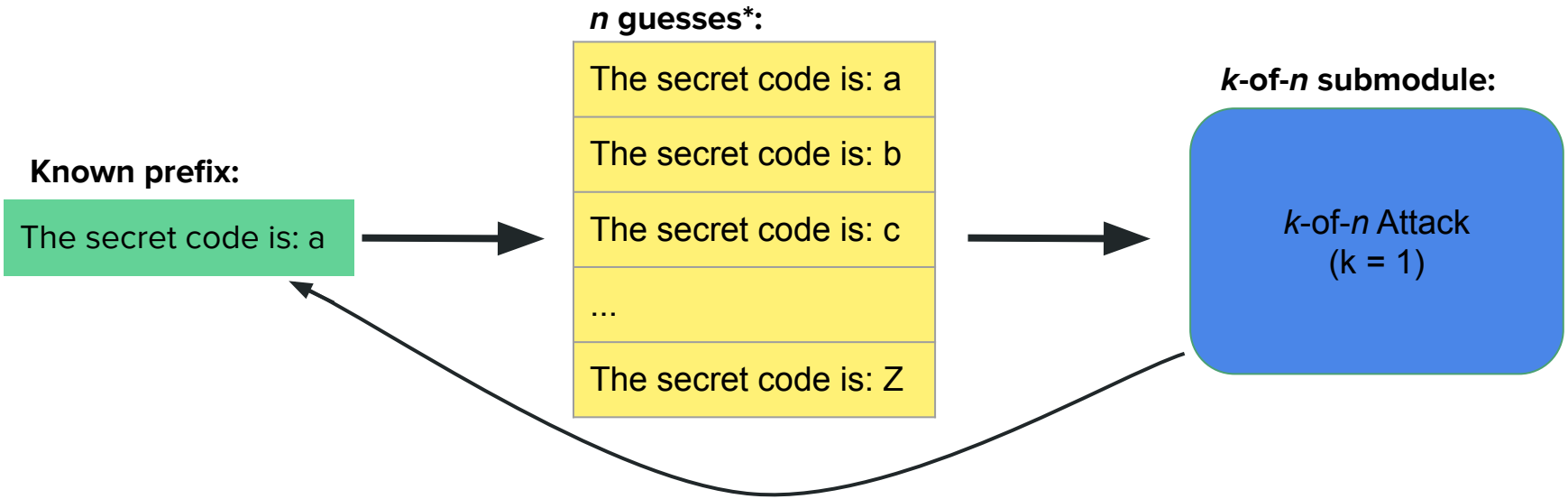

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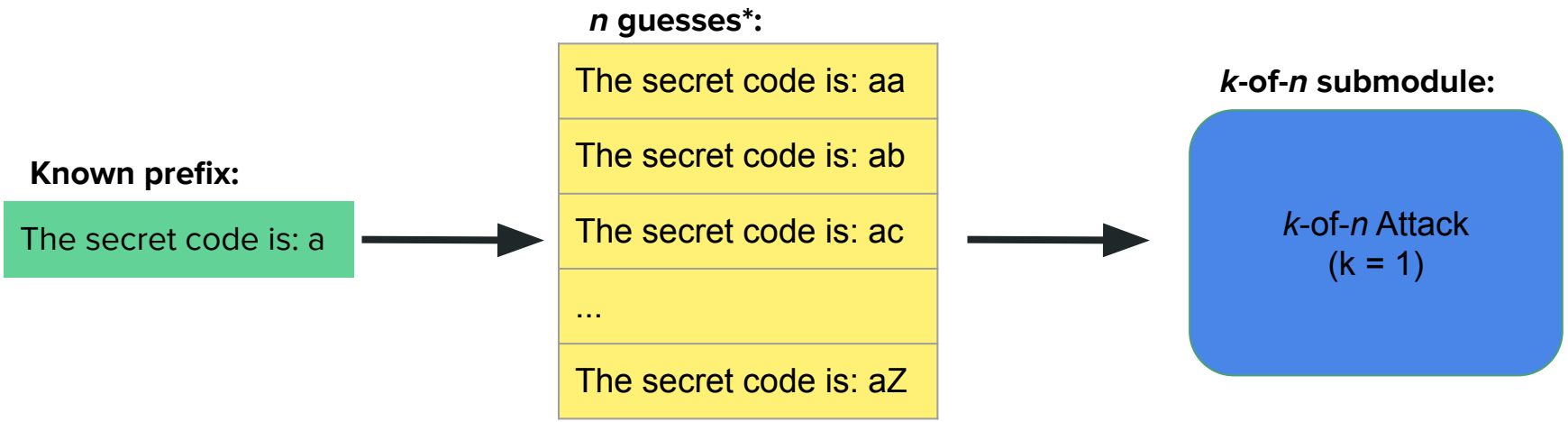

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***n* guesses*:**

- The secret code is: aa
- The secret code is: ab
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- ...
- The secret code is: aZ

***k*-of-*n* submodule:**

k-of-*n* Attack
(k = 1)

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Roadblocks & Optimizations

Substring/Superstring Problem

Might get false positives if guess is a:

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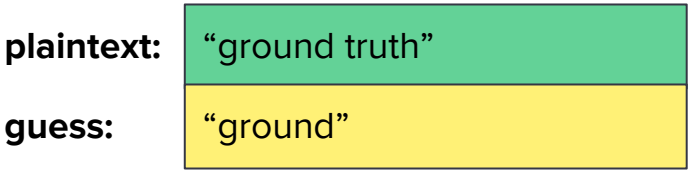

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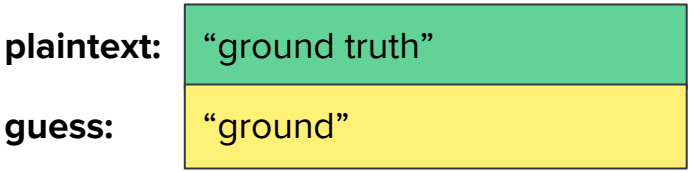

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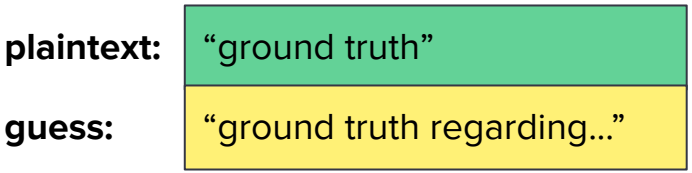

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Still vulnerable to false positives if the superstring is not much longer than the ground truth (recall that we only have to be close to s_{yes} and not precisely match it).

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  - **Solution:** Detect & retry

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Analysis

Efficiency & Speed

After our binary search optimization, the attack becomes very efficient:

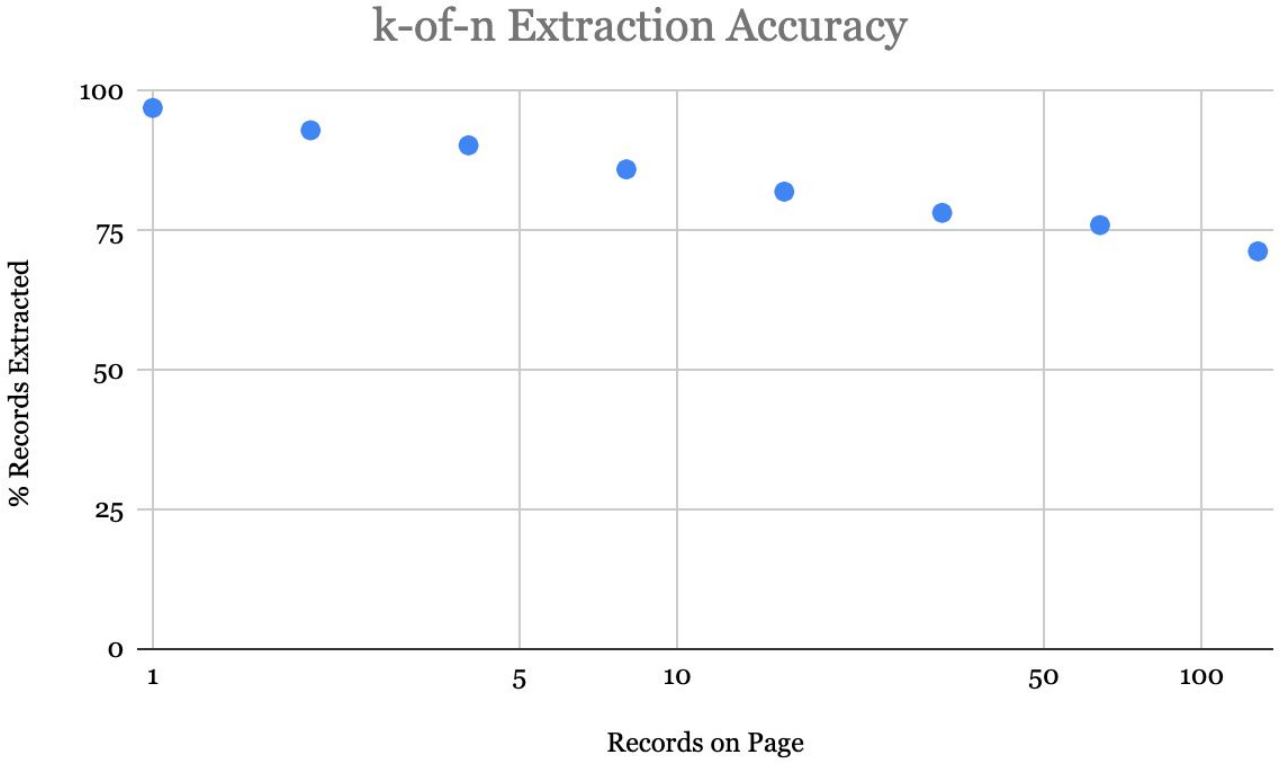
Let *R* be the maximum size of a row

- Insertion of filler rows
  - We must initially insert at most *page_size / R* rows to fill up the page
  - In practice, with an empty page and *R = 200*, this takes about 30 insertions
- Updates per guess
  - $\log_2 R$ updates per guess

Thus, for *n* guesses we perform *O(R + n* log *R)* database actions.

In practice, with *R = 200*, a single guess took 0.2-0.4 seconds.

Accuracy


k-of-n Extraction Accuracy
100
75
50
25
0
% Records Extracted
1
5
10
50
100
Records on Page

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We believe that other RDBMSs and storage engines are vulnerable to the same attack. MySQL is especially likely to be vulnerable.

Mitigations

Prevention

Recommendations for database administrators & developers using databases:

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Only foolproof solution: **Turn off compression.**

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  - Alternatively, require SELECT permissions on all columns in order to UPDATE.
- Compress only within rows
- Or, compress only within rows inserted by the same user / user group

Demo!

DBREACH

- Attack on compression & encryption in databases
- Simple threat model
- **Efficient** and **accurate**

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