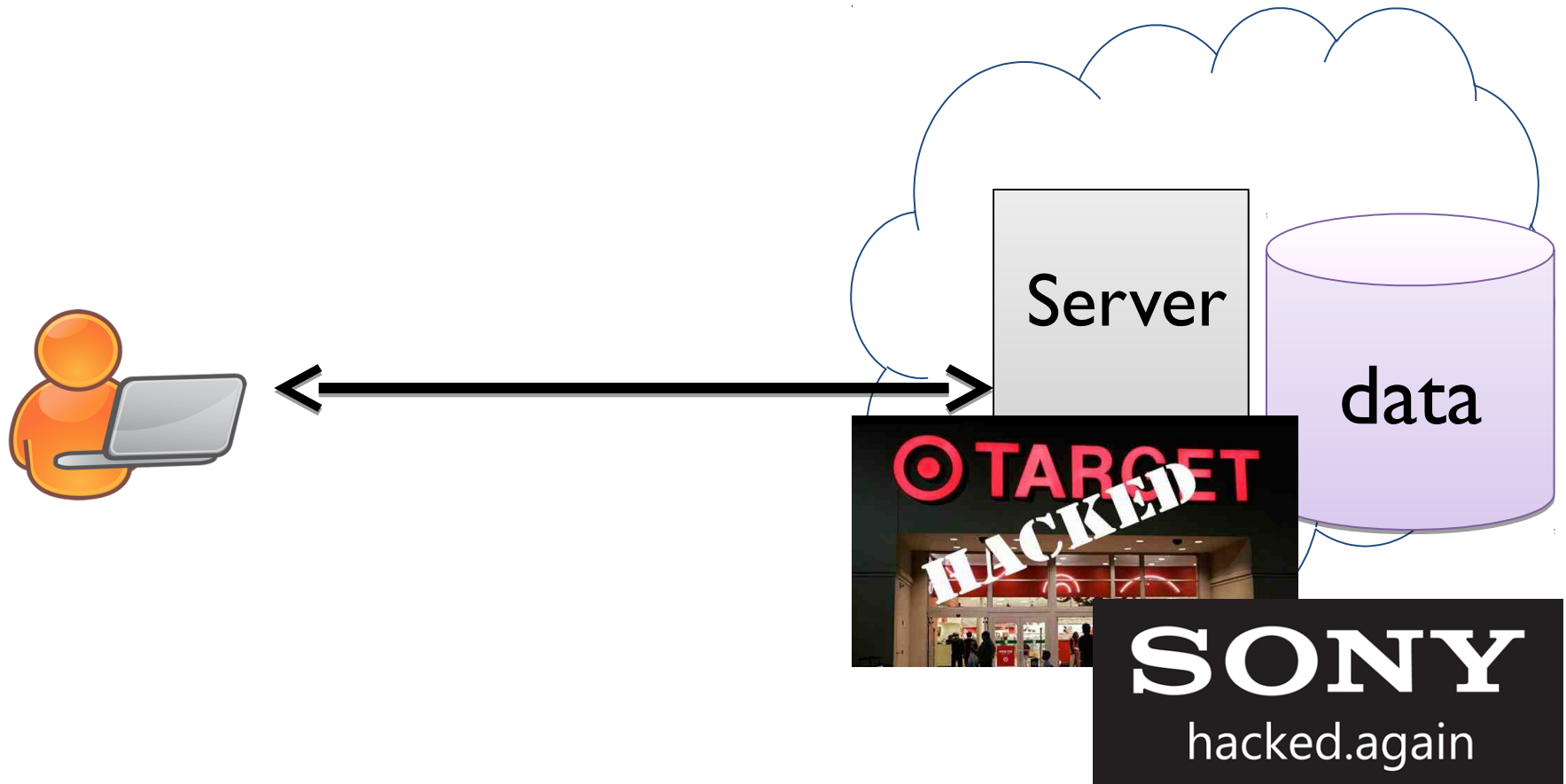
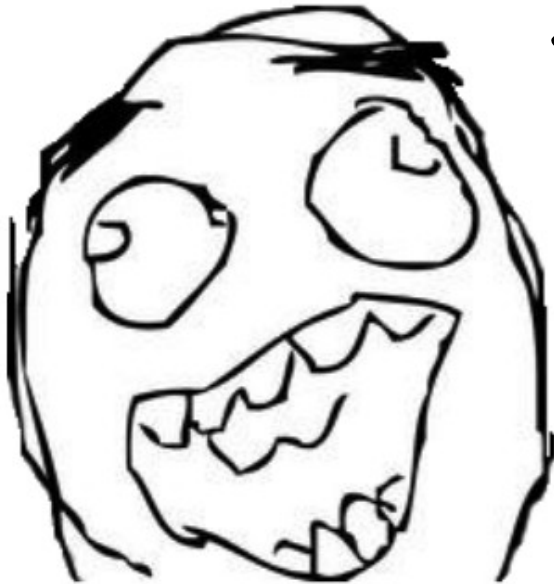


Why Your Encrypted Database Is Not Secure

**Paul Grubbs Tom Ristenpart
Vitaly Shmatikov**

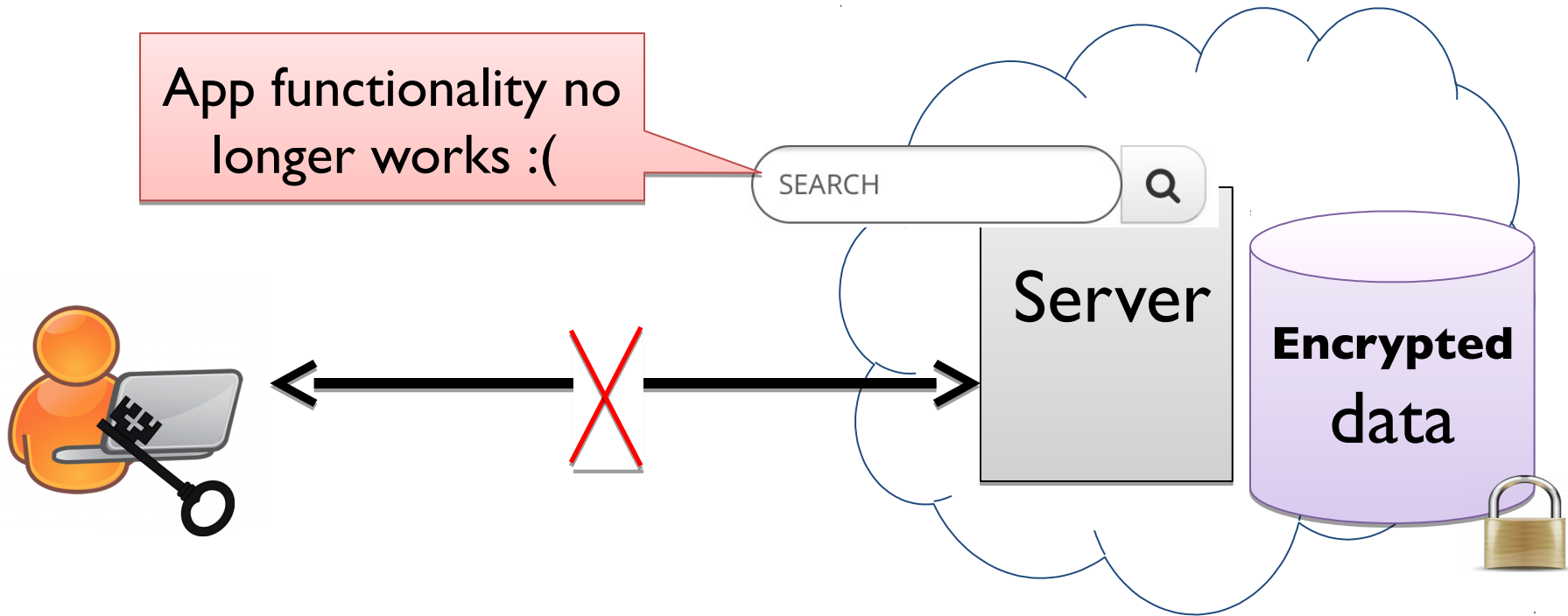
Outsourced Applications Today



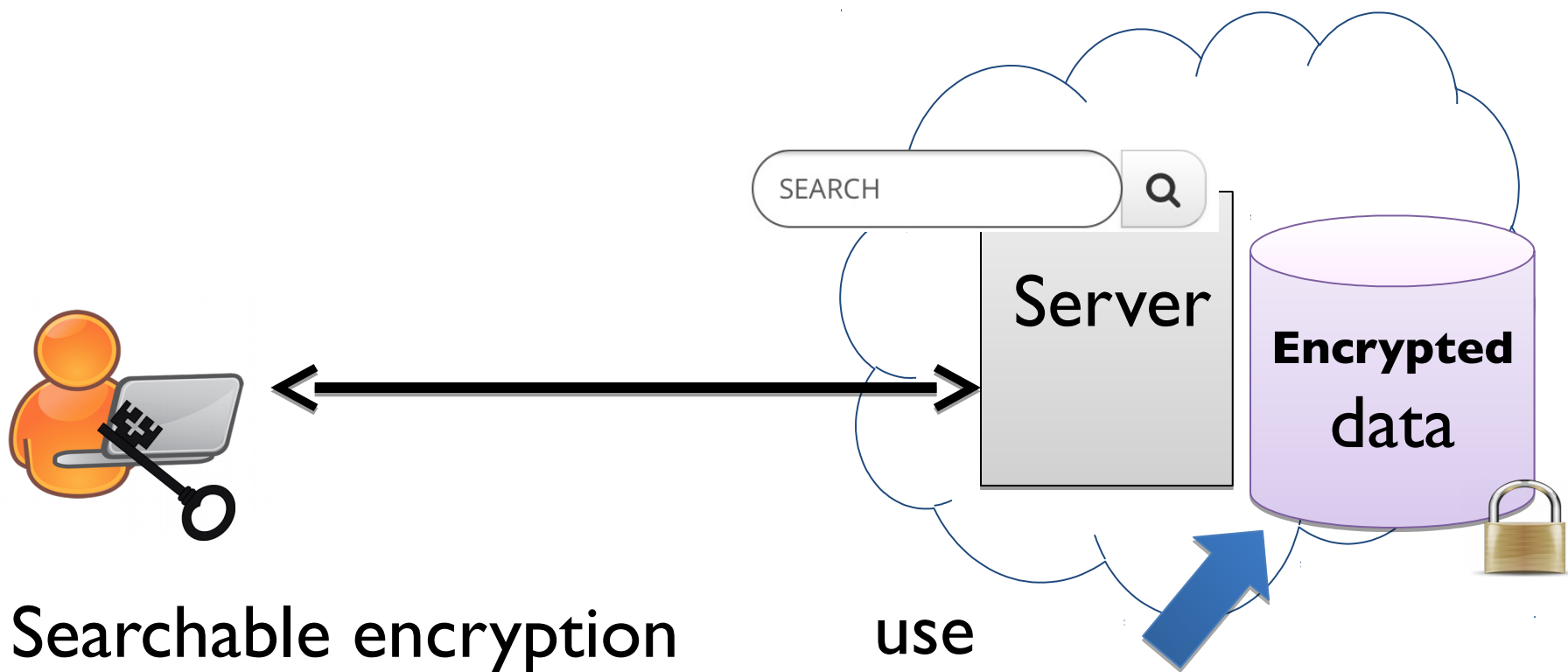


**Encrypt
the data!**

Encrypt the Data



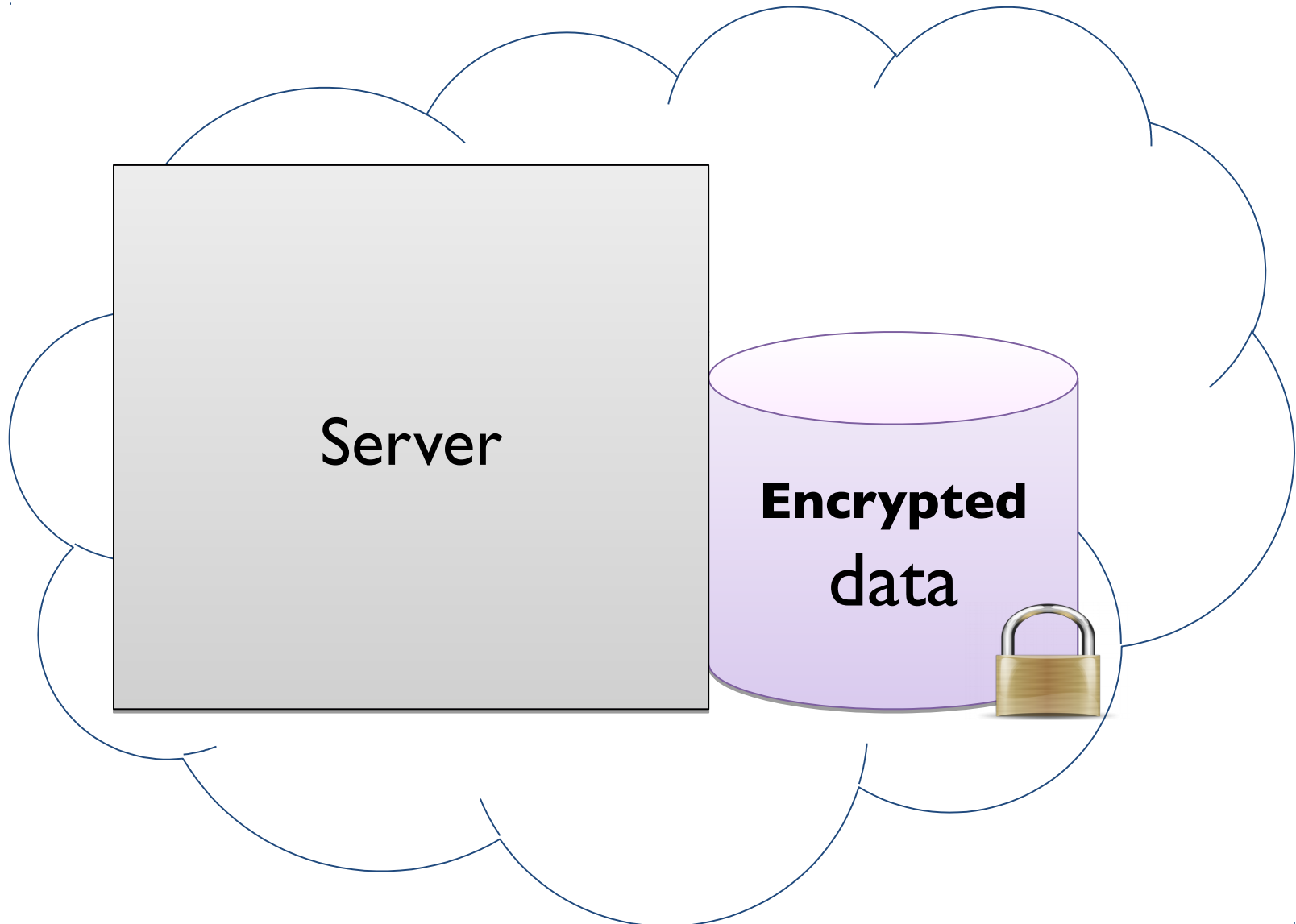
Encrypt the Data



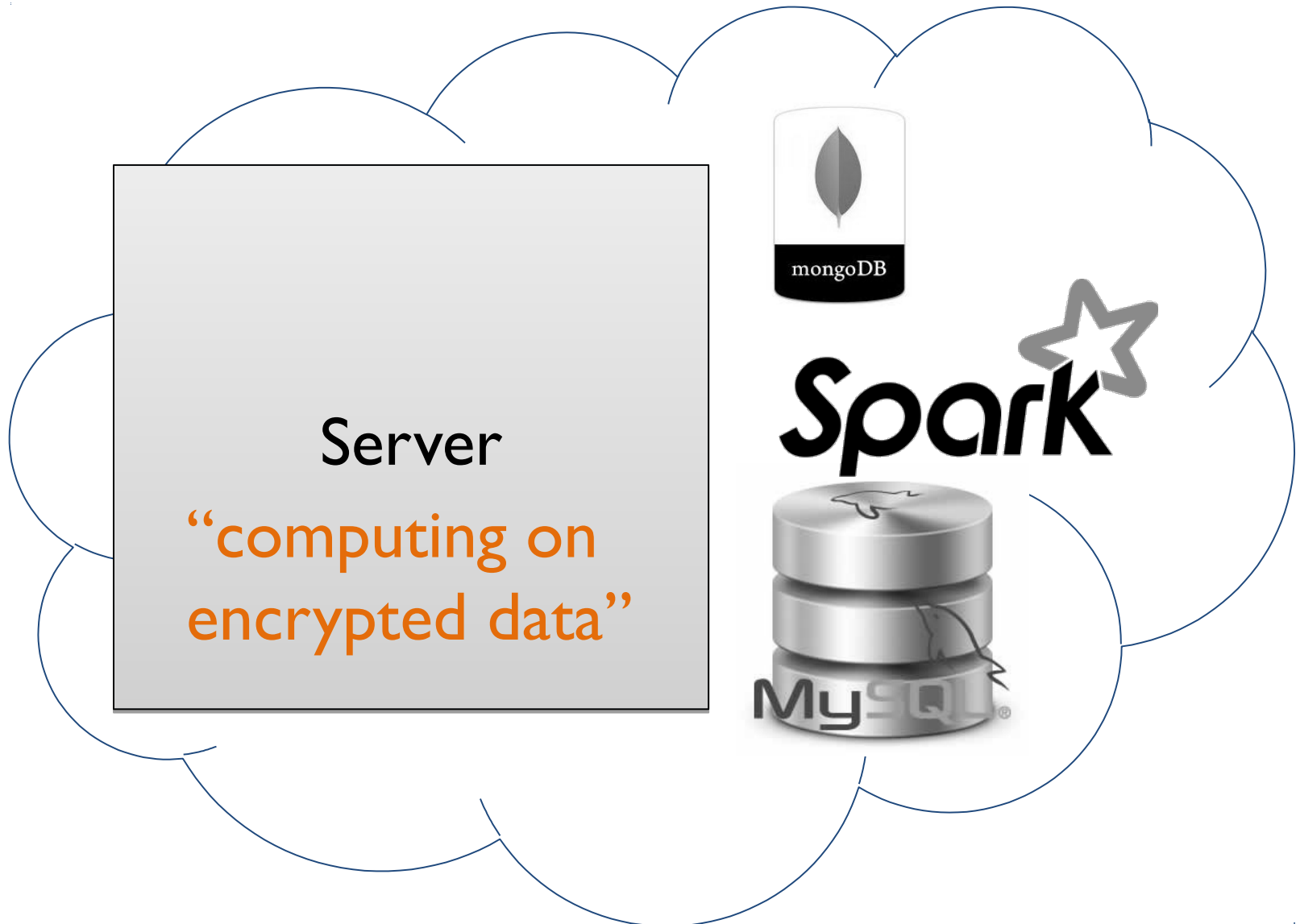
- Searchable encryption
- Deterministic encryption
- Order-revealing encryption

use
property-revealing
encryption (PRE)

Building “Secure” Systems



Building “Secure” Systems



Building “Secure” Systems

- CryptDB (SOSP 2011)
- Mylar (NSDI 2014)
- Seabed (OSDI 2016)
- Arx
- Many others
- Lots of industry and government interest!!



What They Claim

CryptDB is a system that provides practical and provable confidentiality.

Using the “sensitive” annotation, CryptDB ensures that even if an attacker steals an encrypted database, the database does not leak the values of sensitive fields, even if the attacker has side information.

Mylar, a platform for building web applications, which protects data confidentiality against attackers with *full access to servers*.

the server’s encrypted database provides semantic security,

strong security guarantees: it provides an IND-CPA-like security to the database, which reveals nothing beyond sizing information.

“Magically Flexible Cryptography”

CryptDB, on the other hand, manages to emulate fully homomorphic encryption for most of the functions of the SQL databases used in many applications, computing only with encrypted data and adding just 15% to 26% to those applications' computing time.

Claims

em

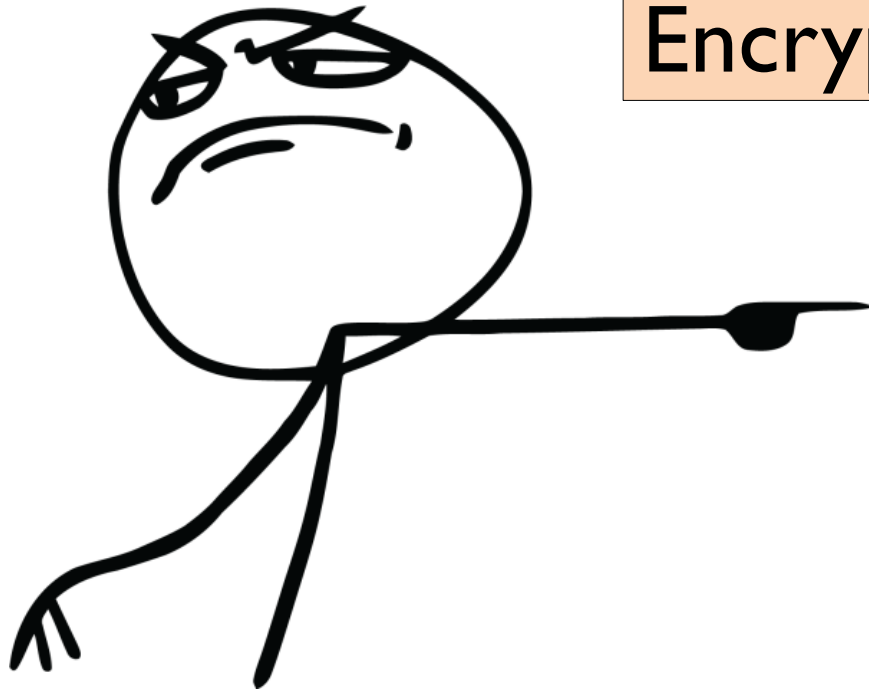
n



the
of

has side information

Fallacy #1



Encryption scheme is “secure”

does not mean

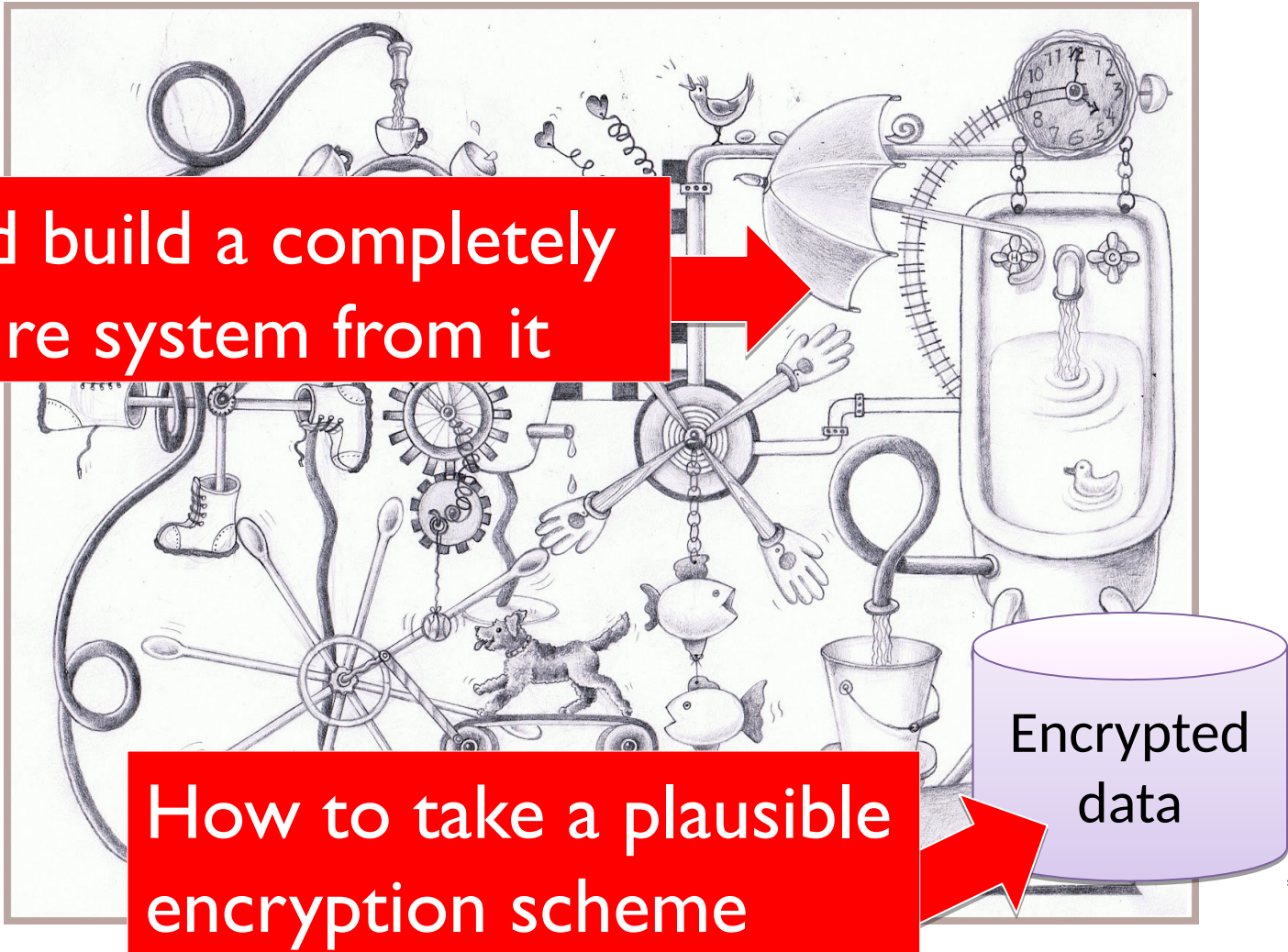
The system is “secure”

What This Talk Is About

... and build a completely insecure system from it

How to take a plausible encryption scheme

Encrypted data



Unsafe at Any Speed

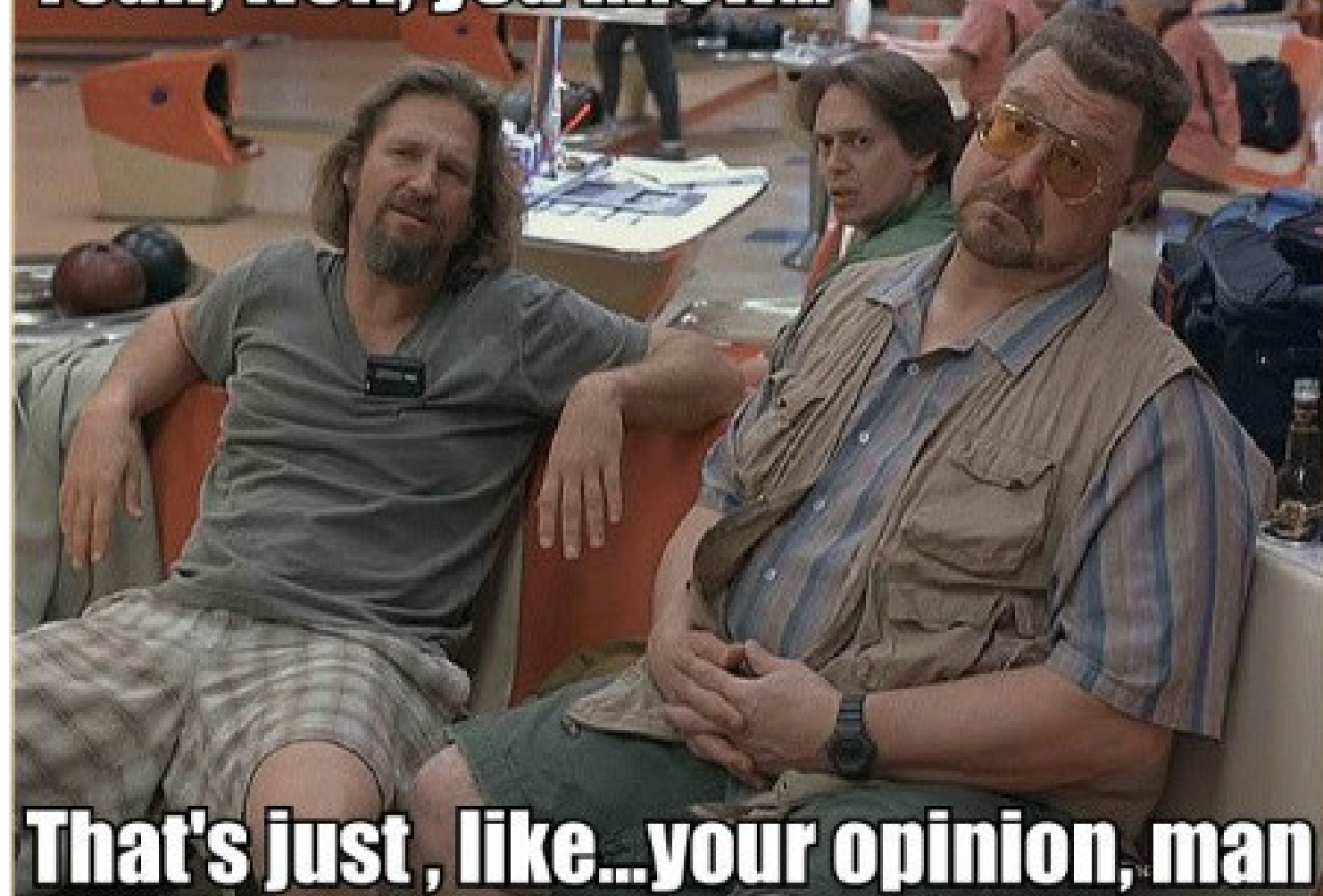
If you look at an actual commodity DBMS ...

- CryptDB (SOSP 2011)
- Mylar (NSDI 2014)
- Seabed (OSDI 2016)
- Arx
- Many others
- Lots of industry and government interest!



... insecure under ANY real-world attack

Yeah, well, you know...



That's just, like...your opinion, man

Threat Models

“Snapshot”



Persistent
passive



Active



Claims Meet Reality

- Secure against active attacks: **false**
 - Grubbs et al. “Breaking web applications built on top of encrypted data” (CCS 2016)
- Secure against “snapshot” attacks: **false**
 - Grubbs et al. “Why your encrypted database is not secure” (HotOS 2017)
- Sensitivity analysis helps: **false**
 - Bindschaedler et al. “The tao of inference in privacy-protected databases” (forthcoming)

Security Against Active Attacks

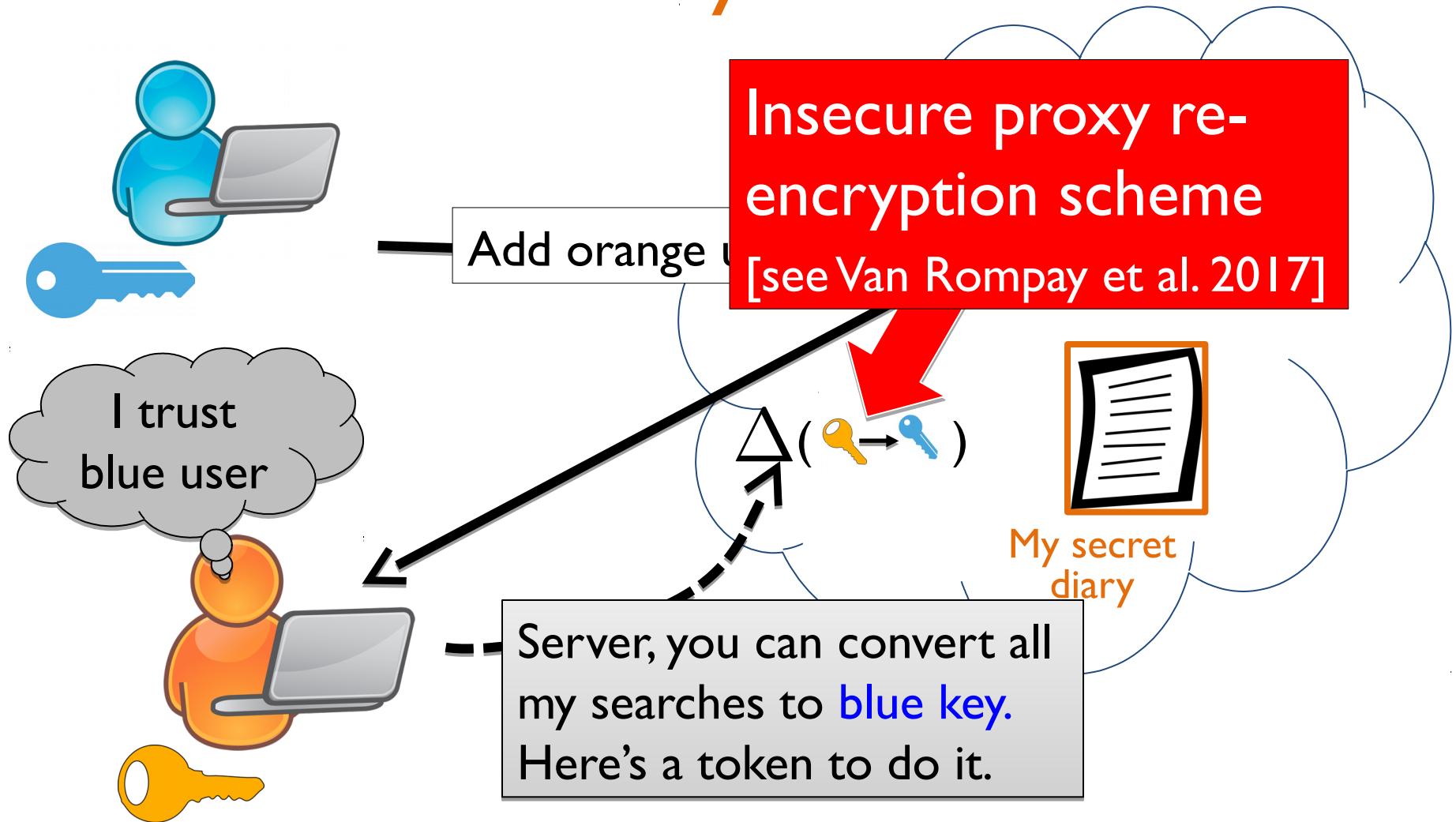
3.4 Threat model

Threats. Both the application and the database servers can be *fully* controlled by an adversary: the adversary may obtain all data from the server, cause the server to send arbitrary responses to web browsers, etc. This model subsumes a wide range of real-world security problems, from bugs in server software to insider attacks.

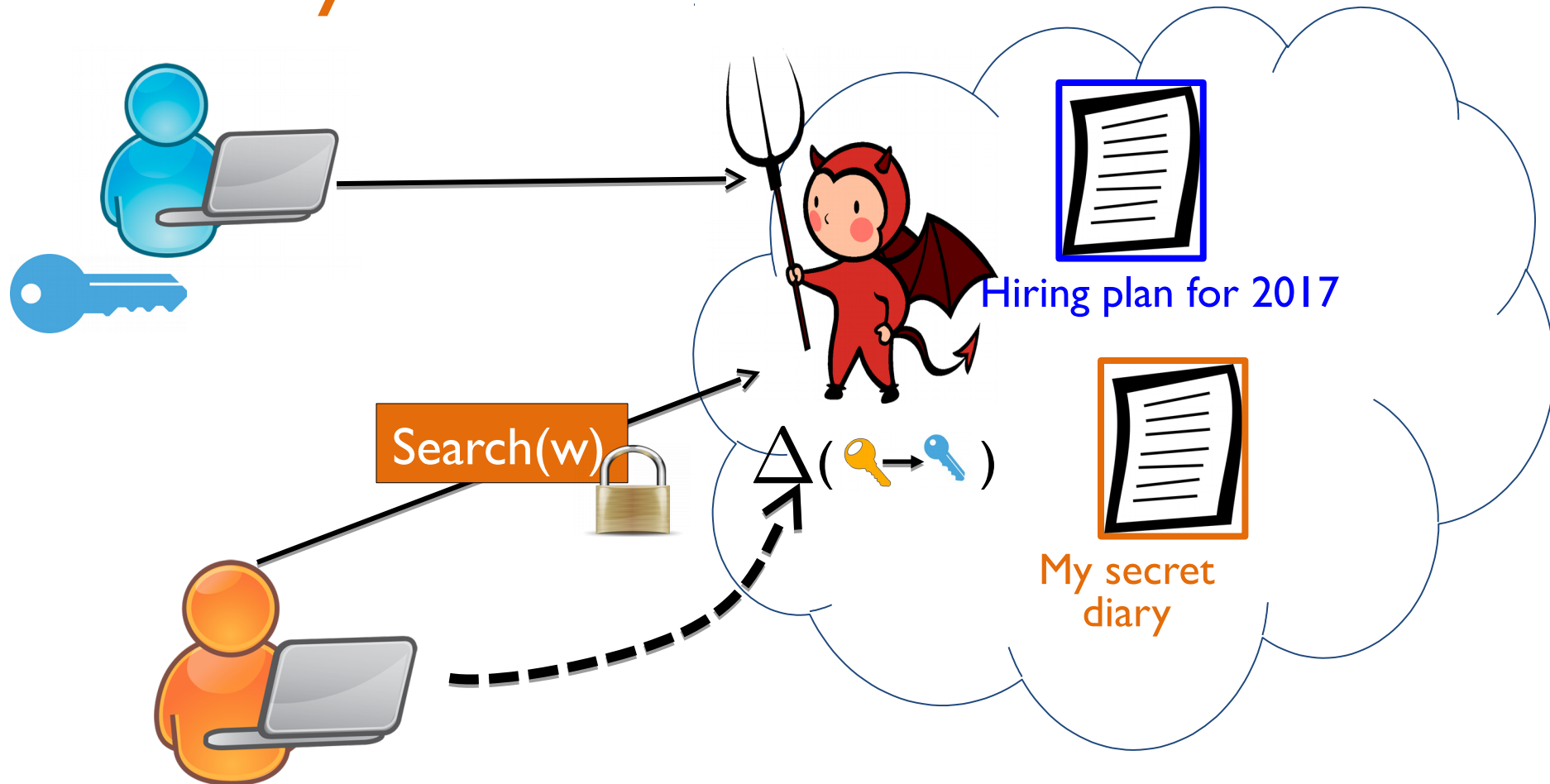
Mylar also allows some user machines to be controlled by the adversary, and to collude with the server. This may be either because the adversary is a user of the application, or because the adversary broke into a user's machine.

We call this adversary *active*, in contrast to a *passive* adversary that eavesdrops on all information at the server, but does not make any changes, so that the server responds to all client requests as if it were not compromised.

Mylar



Mylar Under Active Attack



3.4 Threat model

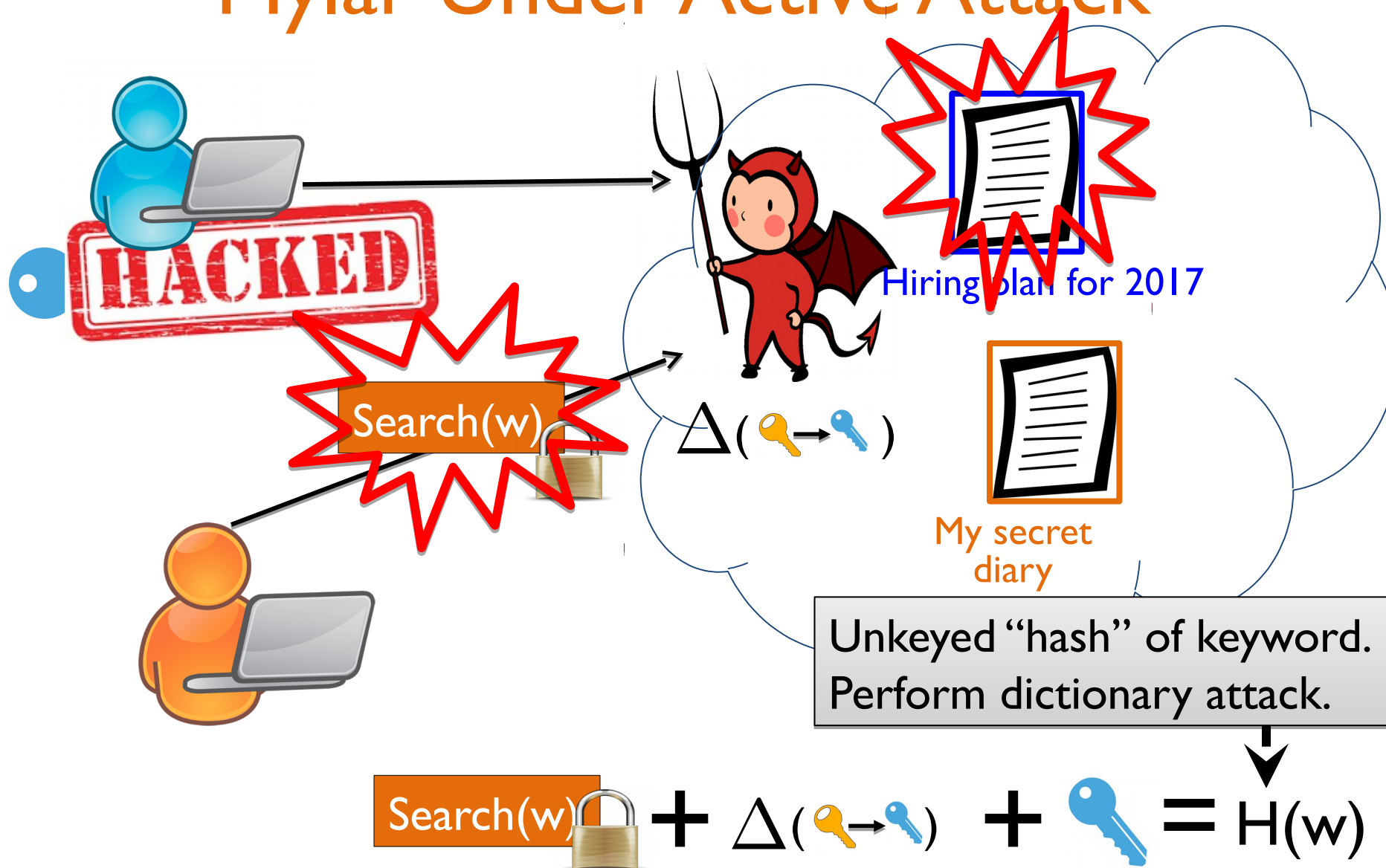
Threats. Both the application and the database servers

can be compromised. Some user machines may collude with the server... **because the adversary broke into a user's machine**

Myiar also allows some user machines to be controlled by the adversary, and to collude with the server. This may be either because the adversary is a user of the application, or because the adversary broke into a user's machine.

We call this adversary *active*, in contrast to a *passive* adversary that eavesdrops on all information at the server, but does not make any changes, so that the server responds to all client requests as if it were not compromised.

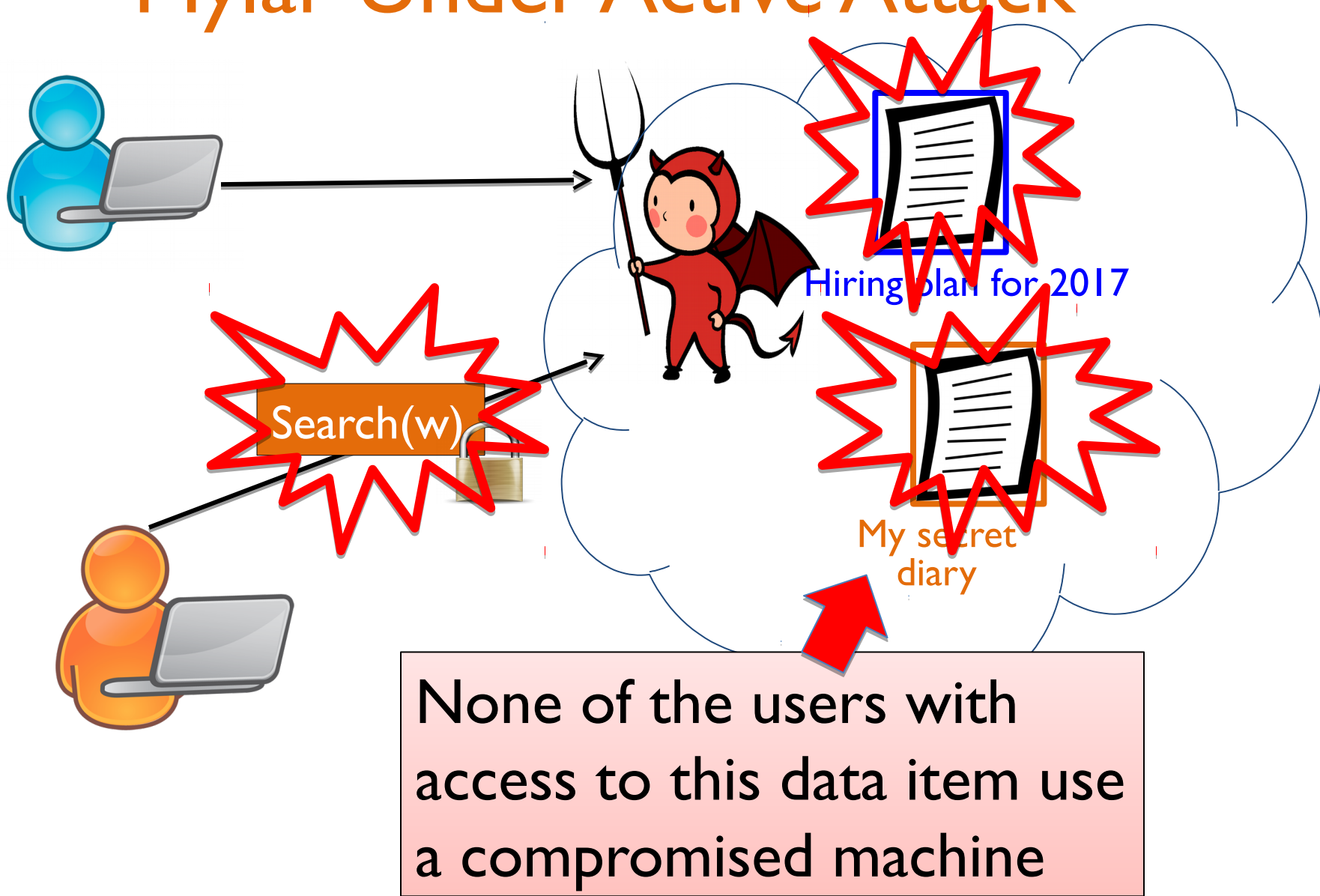
Mylar Under Active Attack



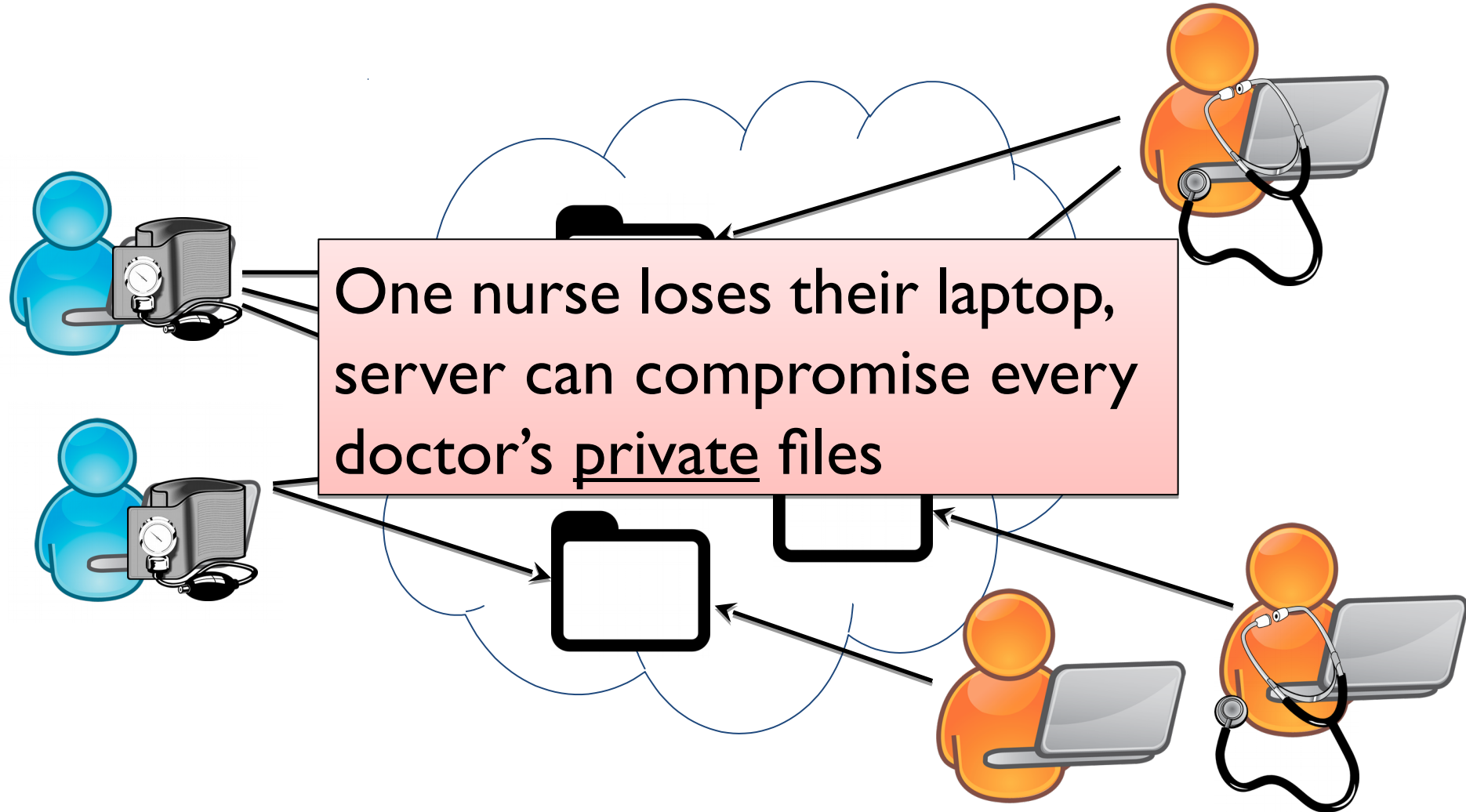
Guarantees. Mylar protects a data item's confidentiality in the face of arbitrary server compromises, as long as none of the users with access to that data item use a compromised machine.

**... as long as none of the
users with access to that
data item use a
compromised machine**

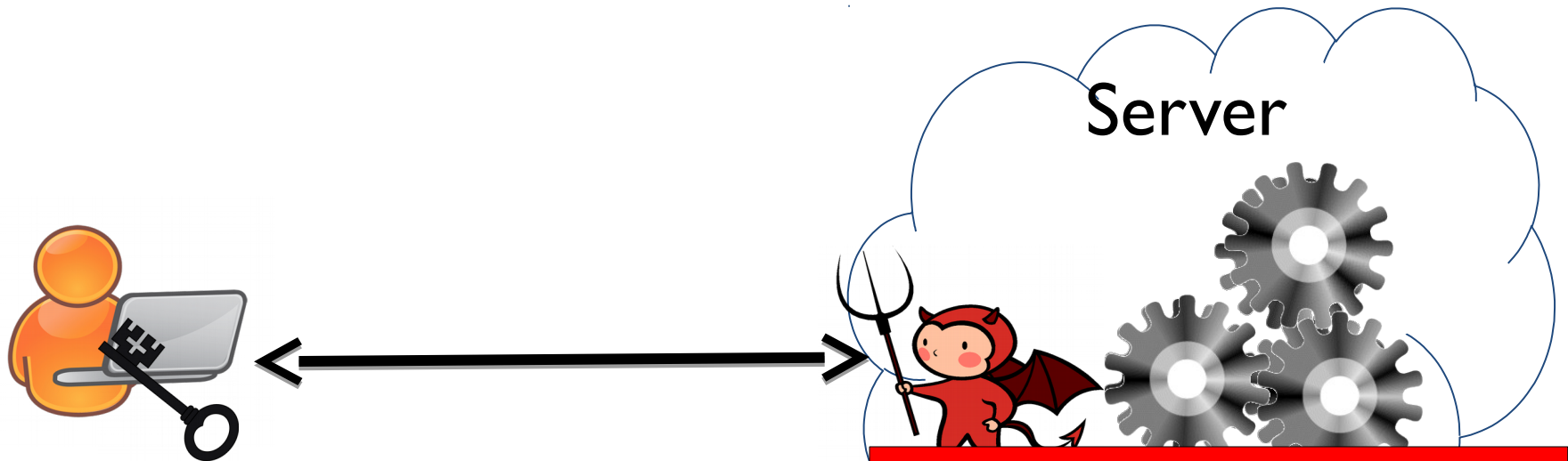
Mylar Under Active Attack



Mylar in a Hospital



“Snapshot” Threat Model

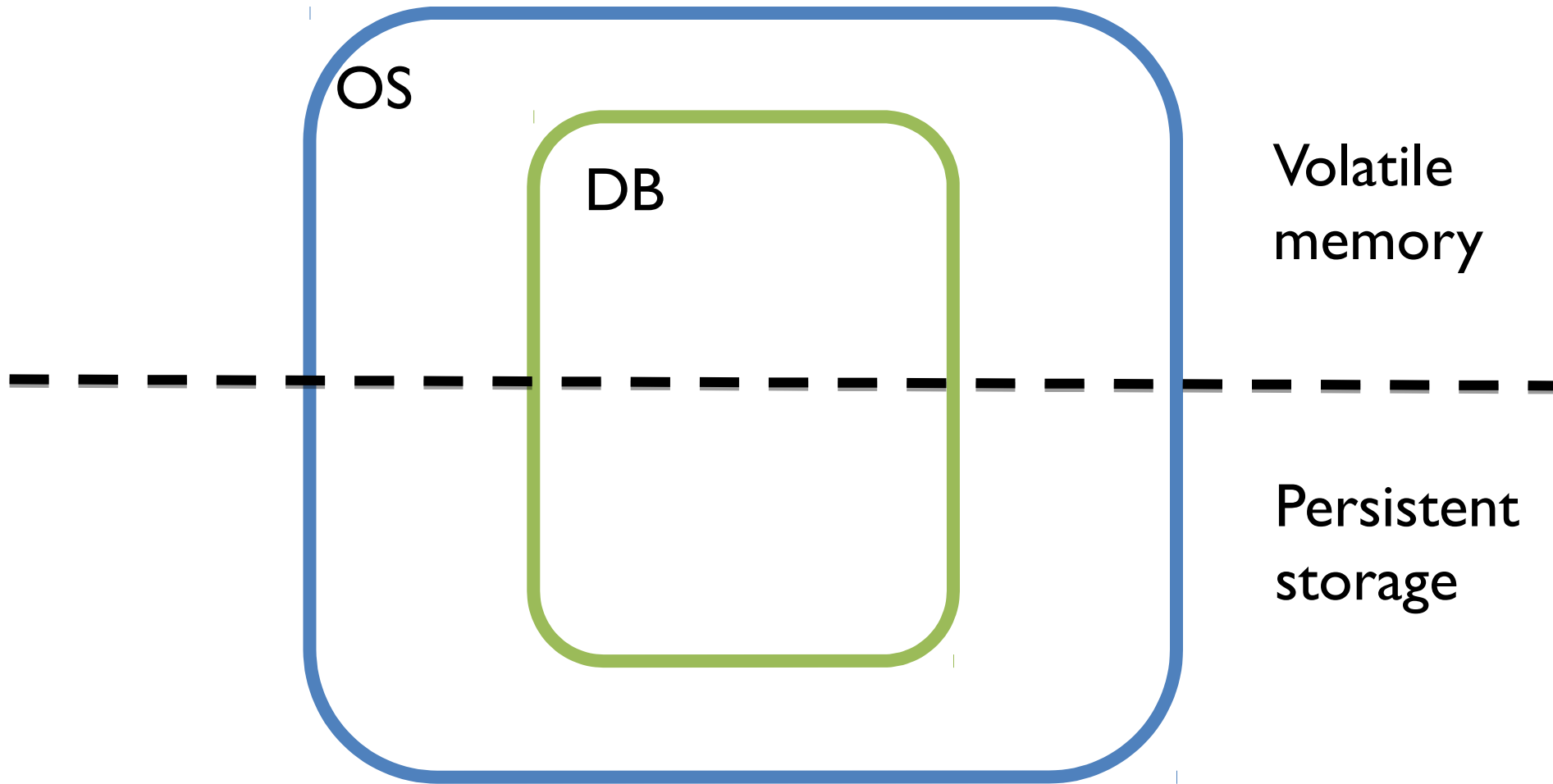


Existing systems explicitly
claim security

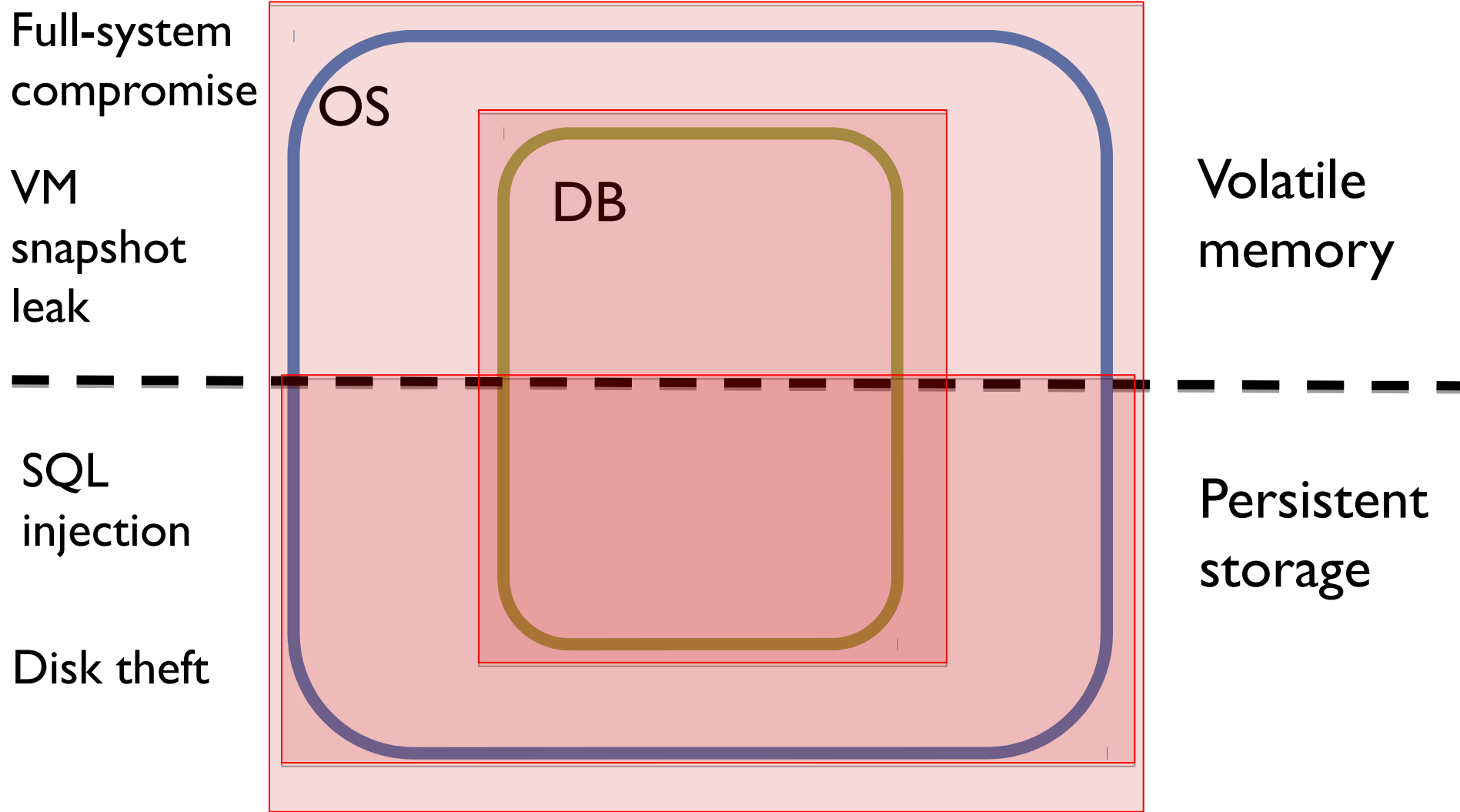
... assuming there are no
queries in the snapshot

False in any realistic
snapshot attack on a
commodity DBMS

A Simple System Abstraction



Actual Attacks



Case Study: MySQL

similar issues in any other commodity DBMS

| Attack | What MySQL leaks | Failed encrypted database |
|--|-----------------------|---------------------------|
| Disk theft | MVCC data structures | Arx's range query index |
| SQL Injection | Past query statistics | Seabed's SPLASHE scheme |
| Full system compromise or VM snapshot leak | Text of past queries | CryptDB, Lewi/Wu, etc. |

Disk Theft

Healthcare IT News

Privacy & Security

Stolen laptop leads to notification for 20 patients

If this is your threat model, just use full-disk encryption



> SC US
SC UK

NEWS CYBERCRIME NETWORK SECURITY PRODUCT REVIEWS IN DEPTH EVENTS

THE CYBERSECURITY SOURCE

SC Magazine US > Blogs > The Data Breach Blog > Hard drive stolen from Jackson Memorial Hospital

Hard drive stolen from Jackson Memorial Hospital

Logs on Disk

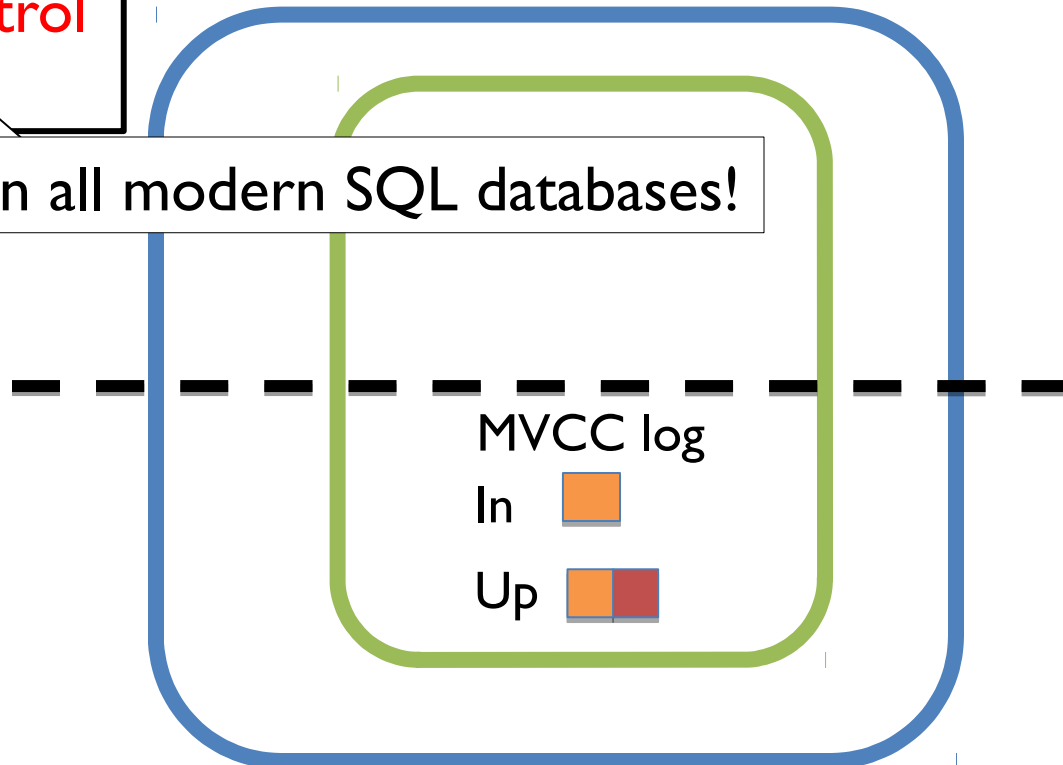
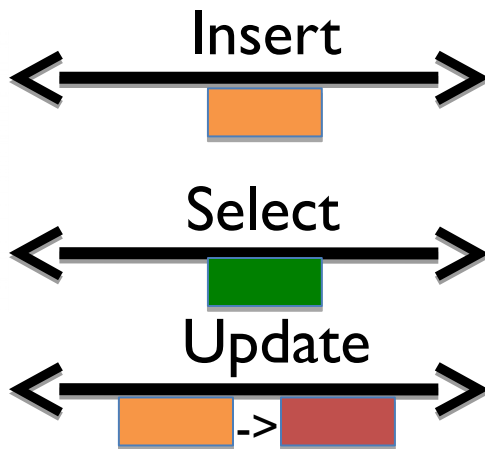
General query log (not widely used)

Binary log records modifications,
used for replication and recovery

Data modification queries
can be reconstructed from
these logs
[FHMW '10, FKSHW '12]

Multi-version concurrency control
using log data structures

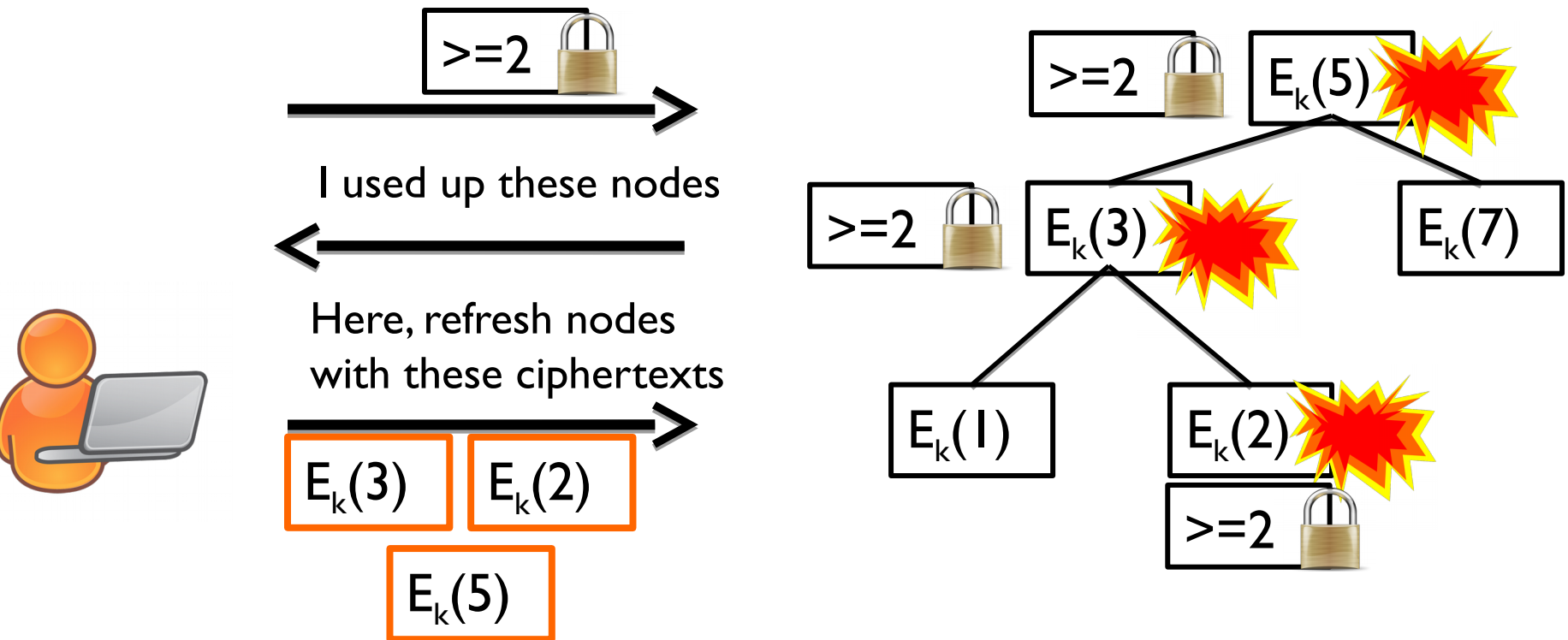
In all modern SQL databases!



Arx

Range queries via chained garbled circuits

Tree nodes become consumed, need replacing



Security Claim for Arx

FALSE

“Arx protects the database with the same level of security as regular AES-based encryption”

Arx Under Snapshot Attack

Range queries via chained garbled circuits

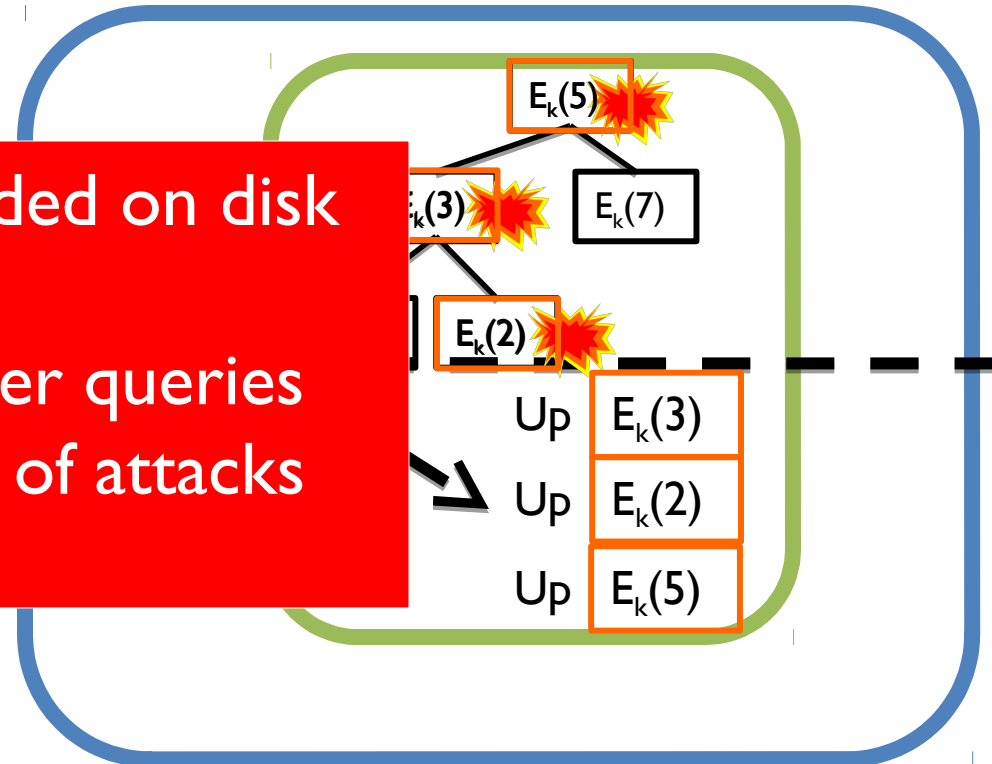
Tree nodes become consumed, need replacing

Consumed nodes immediately replaced,
stored in MVCC log

Query access pattern recorded on disk

Snapshot attacker can recover queries
and plaintexts using variants of attacks
from [GSBNR - S&P '17]

$E_k(5)$

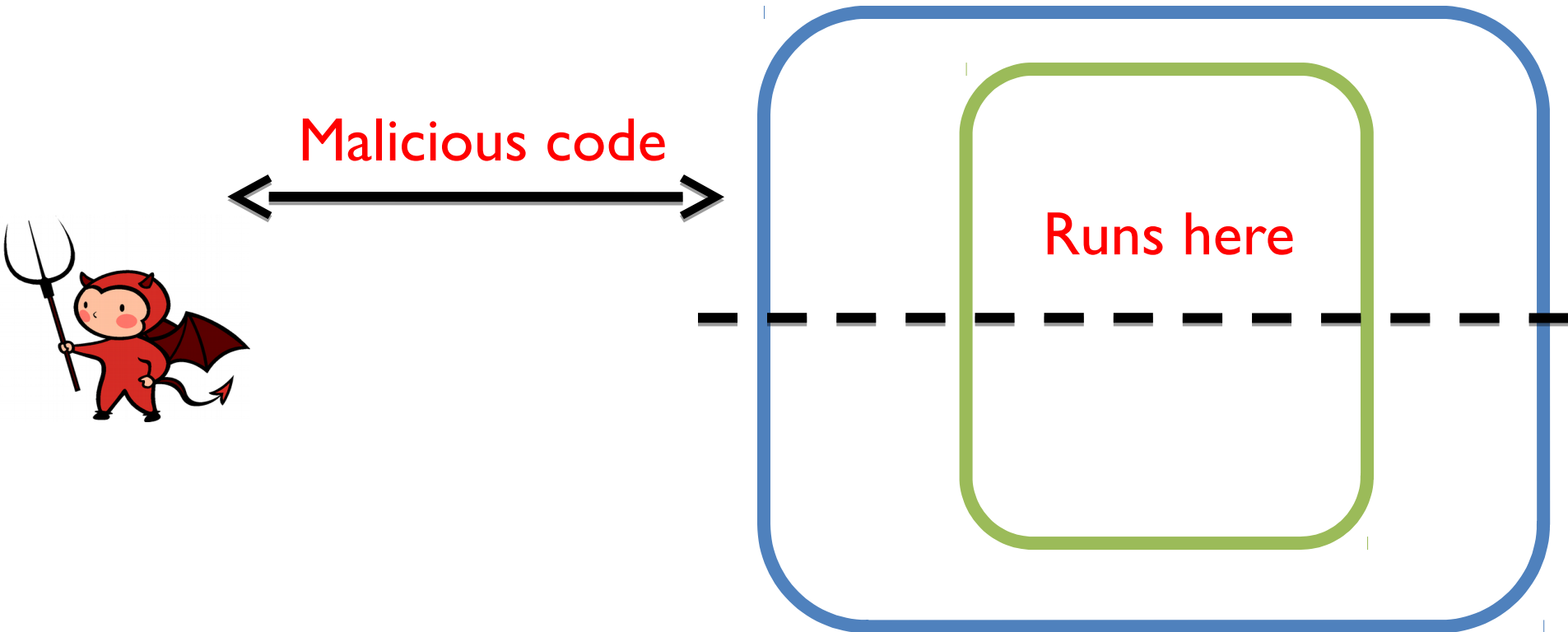


SQL Injection

| Attack | What MySQL leaks | Failed encrypted database |
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SQL Injection

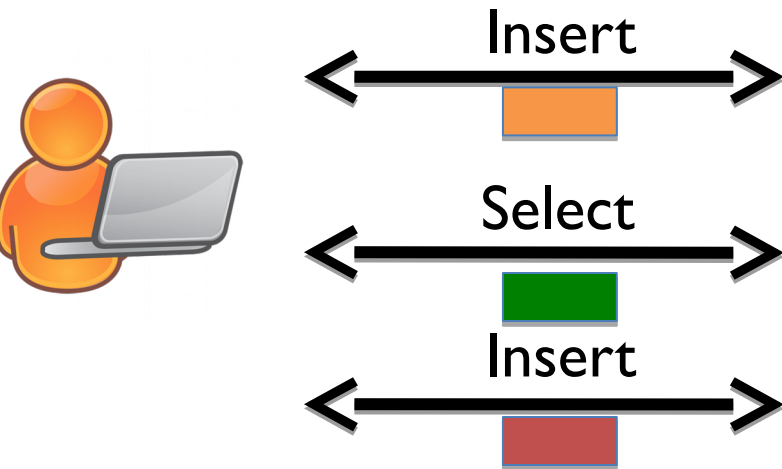
SQL injection accounted for 51% of all Web application attacks in 2016 (source: Akamai)



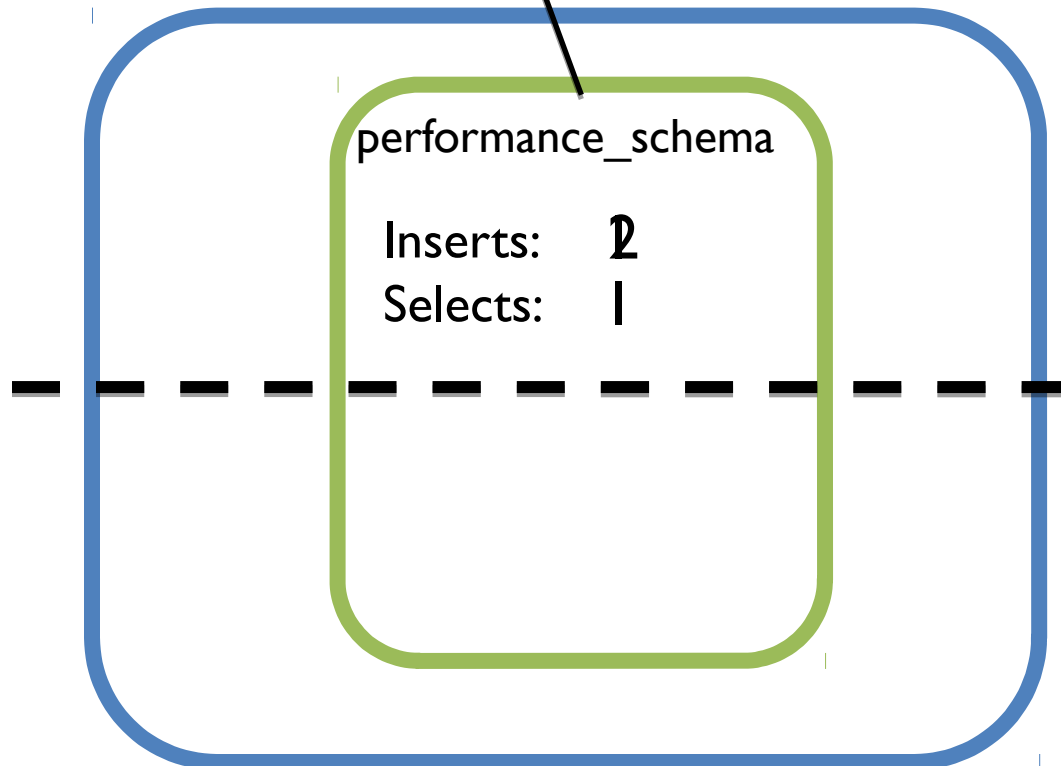
Diagnostic Tables

`information_schema` stores
current query for all users,
contents of buffer cache

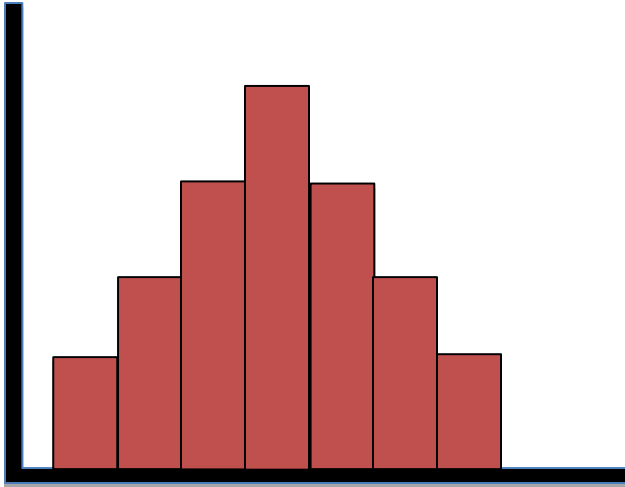
`performance_schema` stores
current query for all threads,
statistics for past queries



Separate counts for queries
which involve different columns



Problem: Frequency Analysis

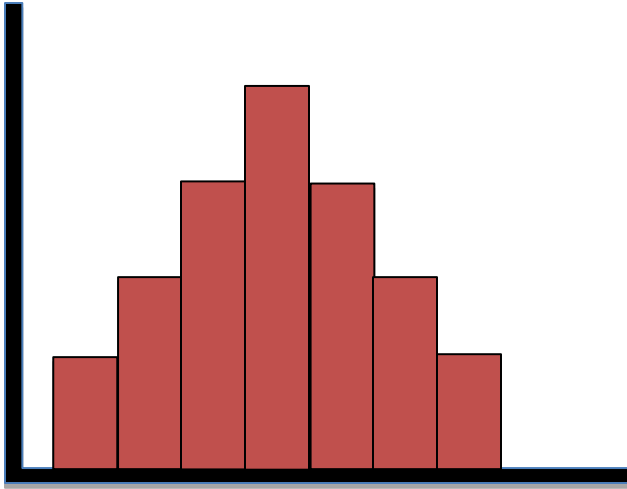


| Name | Has given this talk before |
|-------------------|----------------------------|
| Paul Grubbs | 1 |
| Thomas Ristenpart | 0 |
| Vitaly Shmatikov | 0 |

Order-preserving encryption reveals **histogram** of plaintext values

This is how Naveed et al. used **frequency analysis** to break CryptDB: match histogram to auxiliary model of data distribution

Seabed



Each possible plaintext gets its own column

WHERE clause transformed to correct column

| Name | Has given this talk before |
|-------------------|----------------------------|
| Paul Grubbs | 1 |
| Thomas Ristenpart | 0 |
| Vitaly Shmatikov | 0 |

(“Has ...”=1)

(“Has ...”=0)

| Name | C2 | C3 |
|----------------|----------|----------|
| aspoiwnpoinio | $E_k(1)$ | $E_k(0)$ |
| petryoiueytiew | $E_k(0)$ | $E_k(1)$ |
| Xncmxncmbcn | $E_k(0)$ | $E_k(1)$ |

SELECT Count(“Has ...”) WHERE “Has ...”=1 → SELECT Count(C2)

Separate counts for queries which involve different columns

Example

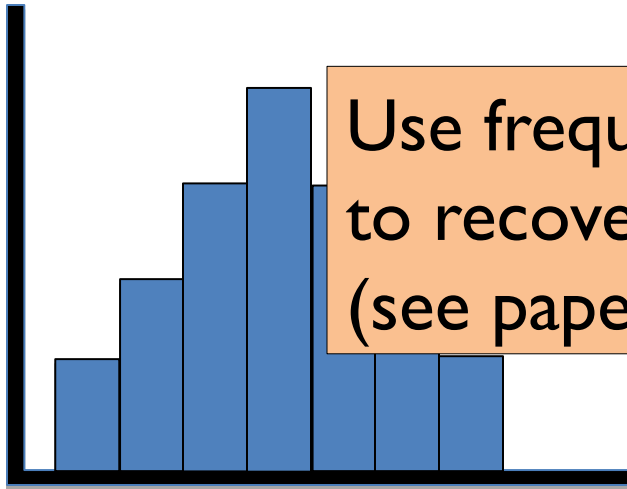
Plaintext Schema

| country | salary |
|---------|--------|
| USA | 100000 |
| USA | 100000 |
| Canada | 200000 |
| USA | 300000 |
| Canada | 500000 |
| Canada | 800000 |
| India | 100000 |
| India | 100000 |
| Chile | 200000 |
| Iraq | 300000 |
| China | 500000 |
| Japan | 800000 |
| Israel | 130000 |
| U.K. | 210000 |

Schema with Enhanced SPLASHE

| country | salaryUSA | salaryCanada | salaryOthers |
|-------------|--------------|--------------|--------------|
| DET(Chile) | ASHE(100000) | ASHE(0) | ASHE(0) |
| DET(Iraq) | ASHE(100000) | ASHE(0) | ASHE(0) |
| DET(China) | ASHE(0) | ASHE(200000) | ASHE(0) |
| DET(Japan) | ASHE(300000) | ASHE(0) | ASHE(0) |
| DET(Israel) | ASHE(0) | ASHE(500000) | ASHE(0) |
| DET(U.K.) | ASHE(0) | ASHE(800000) | ASHE(0) |
| DET(India) | ASHE(0) | ASHE(0) | ASHE(100000) |
| DET(India) | ASHE(0) | ASHE(0) | ASHE(100000) |
| DET(Chile) | ASHE(0) | ASHE(0) | ASHE(200000) |
| DET(Iraq) | ASHE(0) | ASHE(0) | ASHE(300000) |
| DET(China) | ASHE(0) | ASHE(0) | ASHE(500000) |
| DET(Japan) | ASHE(0) | ASHE(0) | ASHE(800000) |
| DET(Israel) | ASHE(0) | ASHE(0) | ASHE(130000) |
| DET(U.K) | ASHE(0) | ASHE(0) | ASHE(210000) |

SQLi Extracts Diagnostic Tables



Use frequency analysis
to recover plaintexts
(see paper for details)



SELECT Count(C3)



SELECT Count(C2)



SELECT Count(C3)



performance_schema:

Selects for C2: 1

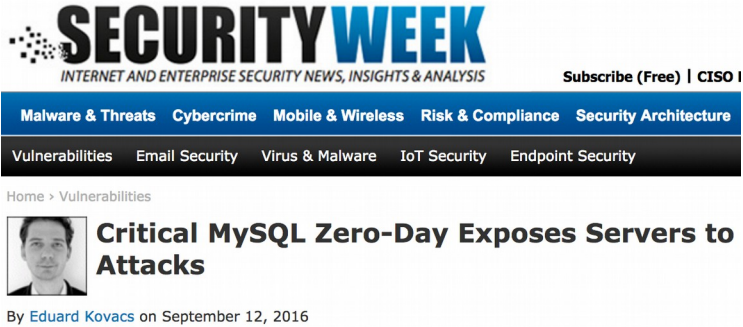
Selects for C3: 2

Separate counts for queries which involve different columns

Full-System Snapshot

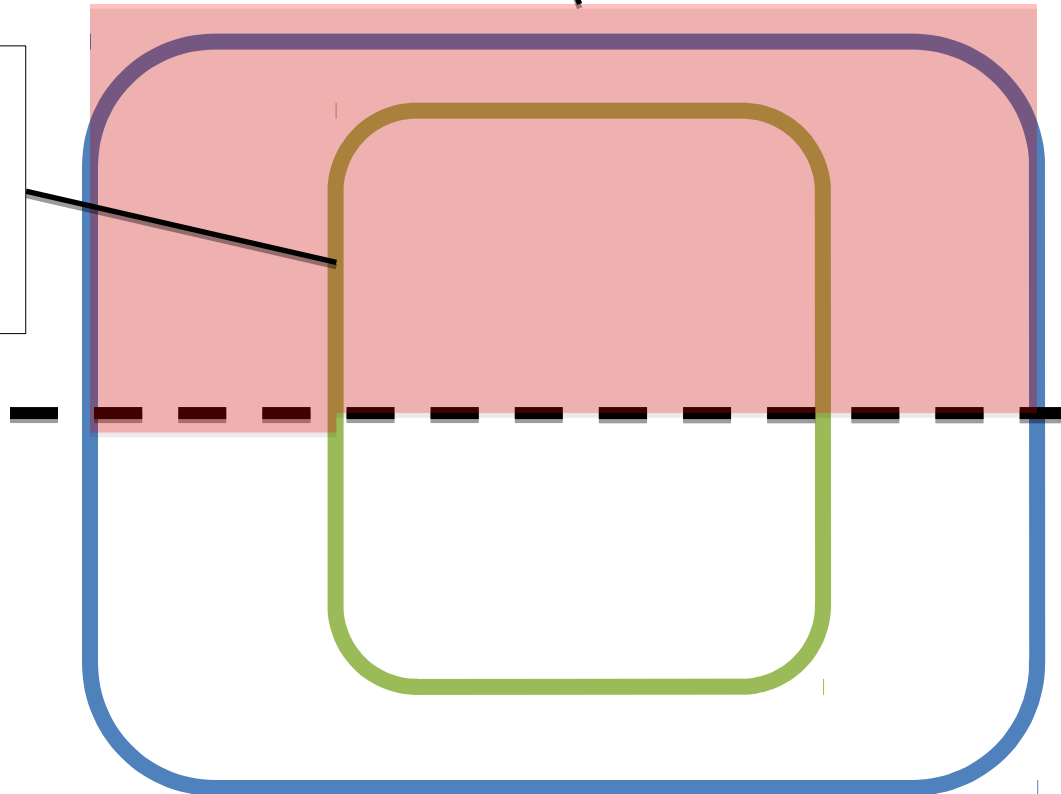
| Attack | What MySQL leaks | Failed encrypted database |
|--|-----------------------|---------------------------|
| Disk theft | MVCC data structures | Arx's range query index |
| SQL Injection | Past query statistics | Seabed's SPLASHE scheme |
| Full system compromise or VM snapshot leak | Text of past queries | CryptDB, Lewi/Wu, etc. |

Full-System Compromise



Leakage of sensitive data at OS level is well-studied
[CPGR, DL]KSXSW

We focus on DBMS
address space, things
inaccessible to users



Data Structures and Caches

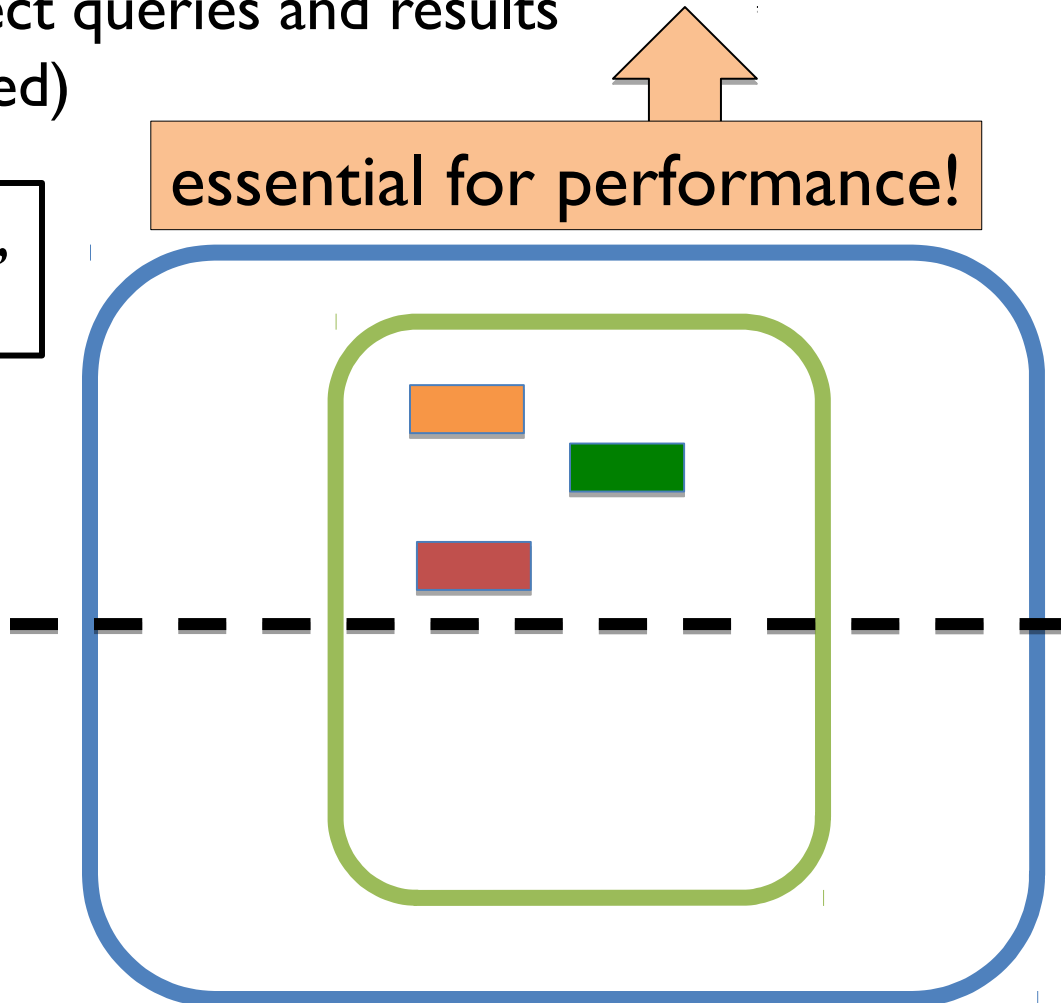
Adaptive hash index tracks pages accesses, indexes automatically

MySQL **query cache** stores select queries and results

Other query caches (memcached)

MySQL manages internal heaps,
does not zero freed memory!

essential for performance!



Token-Based Systems

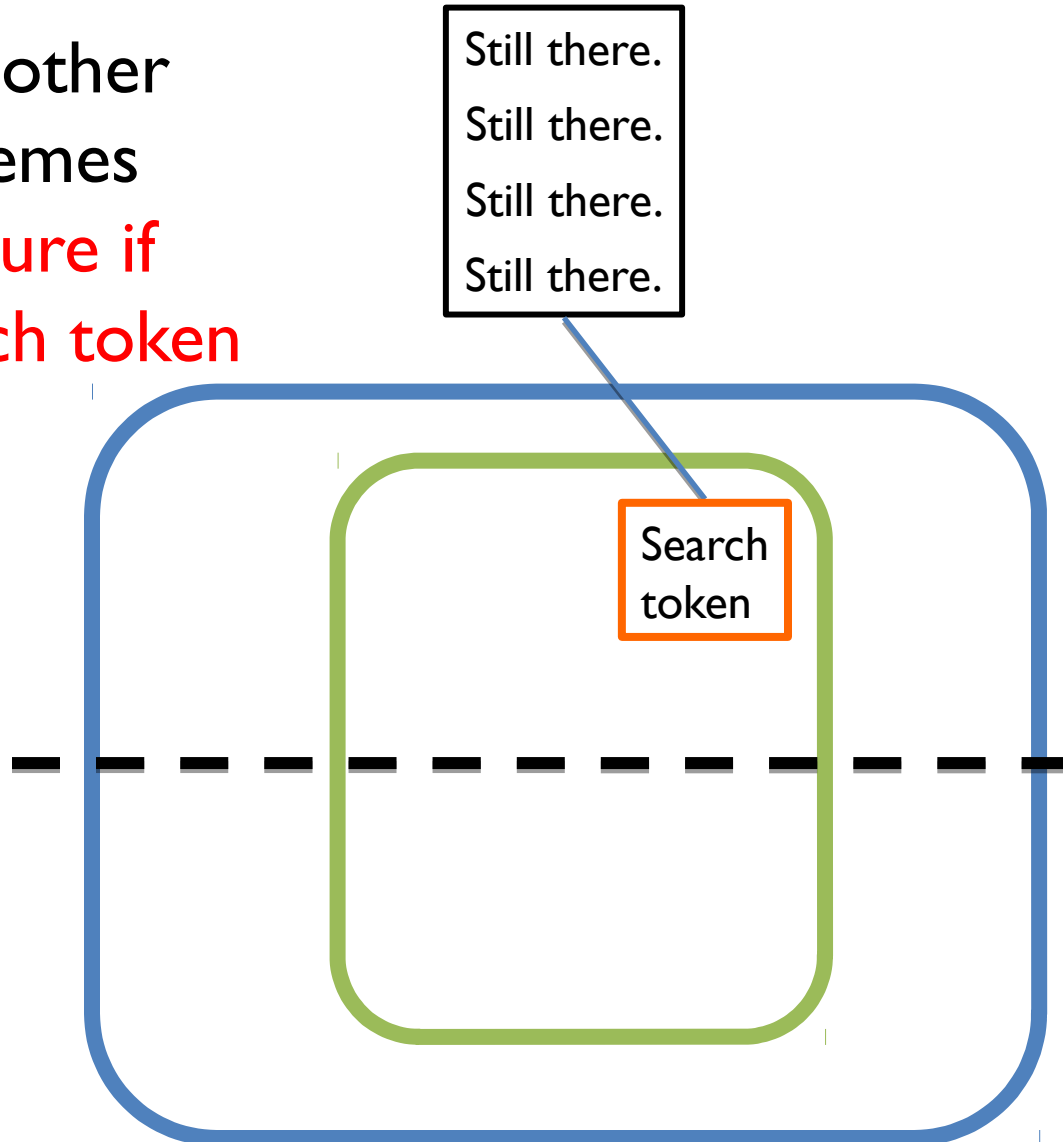
CryptDB, Mylar, Lewi-Wu, other searchable encryption schemes **cannot be semantically secure if attacker sees a single search token**



1,000 random selects...

Waited a while...

100,000 more random selects...



Let Me Make Myself Perfectly Clear

A still from a movie showing Al Pacino in a dark suit, holding a machine gun with a bloodied hand, looking intense. The image is used as a background for the text.

These encrypted databases **CANNOT** be semantically secure under **ANY** real-world attack

There is no such thing as a snapshot attack
There is no such thing as a snapshot attack
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“I Will Build My Own Database”

You can try...

Transaction logs needed to support ACID

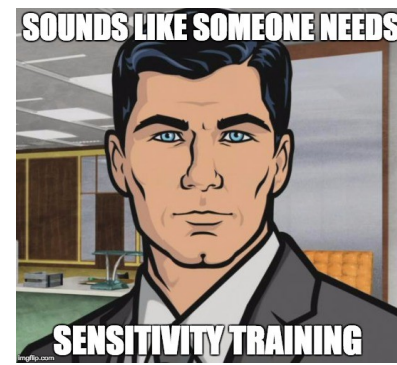
Log-structured storage

Caching

Adaptive data structures adjust to workload

... everything in modern databases leaks
information about past queries

Sensitivity Analysis



| SSN | Name | Ethnicity | Date Of Birth | Sex | ZIP | Marital Status | Problem |
|-----|------|-----------|---------------|--------|-------|----------------|---------------------|
| | | black | 09/27/64 | male | 02139 | divorced | obesity |
| | | black | 09/30/64 | male | 02139 | divorced | hypertension |
| | | black | 04/18/64 | male | 02139 | married | chest pain |
| | | black | 04/15/64 | male | 02139 | married | chest pain |
| | | black | 09/15/64 | male | 02138 | married | shortness of breath |
| | | caucasian | 03/13/63 | male | 02141 | married | hypertension |
| | | caucasian | 03/18/63 | male | 02141 | married | shortness of breath |
| | | caucasian | 09/13/64 | female | 02138 | married | shortness of breath |
| | | caucasian | 09/07/64 | female | 02138 | married | obesity |
| | | caucasian | 05/14/61 | female | 02138 | single | chest pain |
| | | caucasian | 05/08/61 | female | 02138 | single | obesity |

order-preserving
encryption



can sort

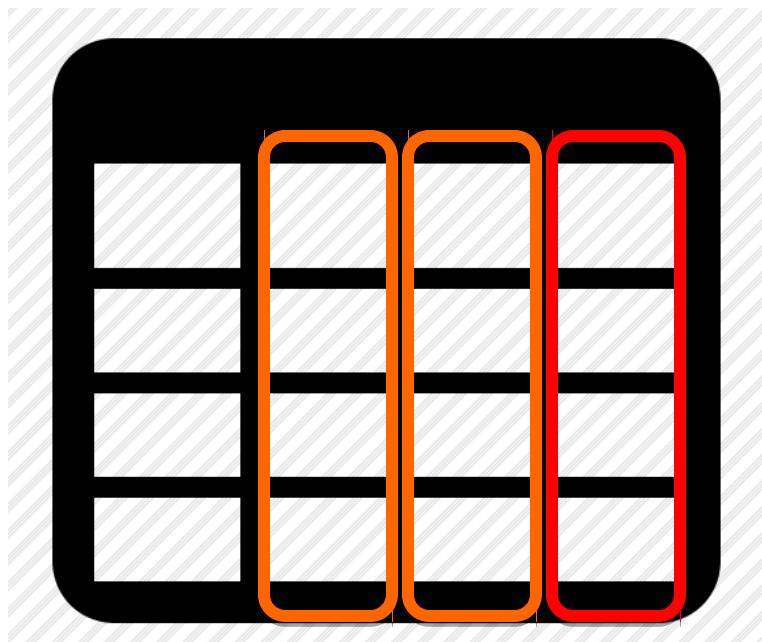
deterministic
encryption



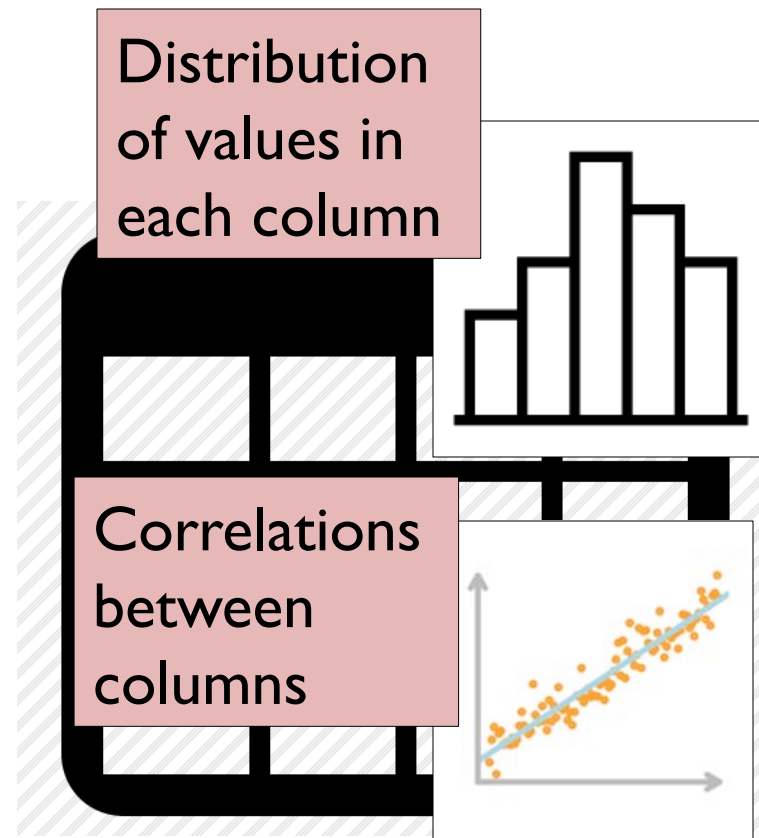
can check for equality

“strong”
encryption

Auxiliary Data

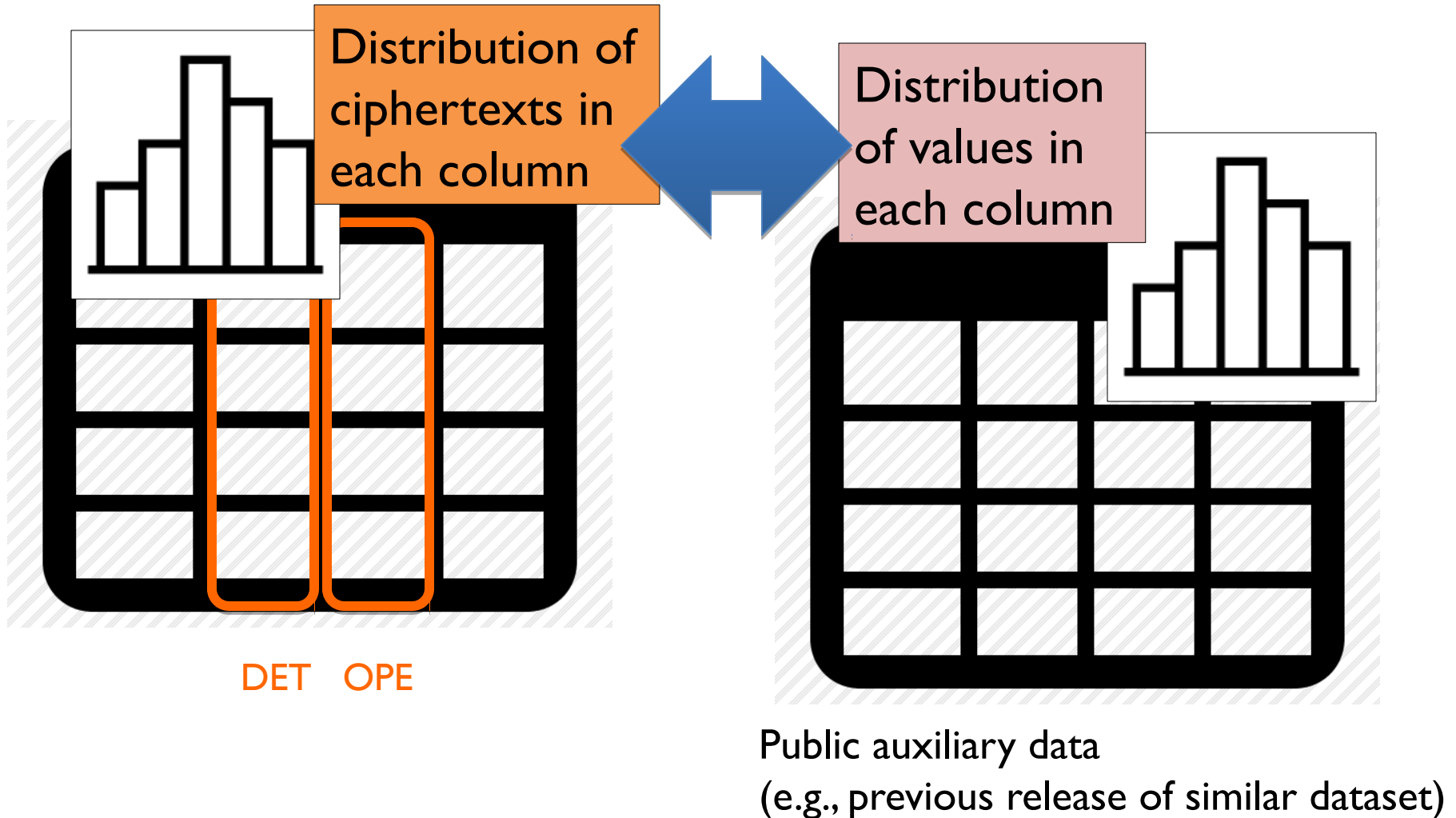


DET OPE IND-CPA



Public auxiliary data
(e.g., previous release of similar datasets)

Bayesian Inference



Multinomial Attack

Observed
ciphertexts

Plaintext distribution
(from auxiliary data)

$$\Pr\{\mathbf{f} = f \mid \vec{c}; \rho\} = \frac{\Pr\{\vec{c} \mid \mathbf{f} = f; \rho\} \cdot \Pr\{\mathbf{f} = f; \rho\}}{\Pr\{\vec{c}; \rho\}}$$

$$f_{\max} = \arg \max_f \Pr\{\mathbf{f} = f \mid \vec{c}; \rho\}$$



Most likely
mapping of
ciphertexts
to plaintexts

$$= \arg \max_f \Pr\{\vec{c} \mid \mathbf{f} = f; \rho\} \cdot \Pr\{\mathbf{f} = f; \rho\}$$

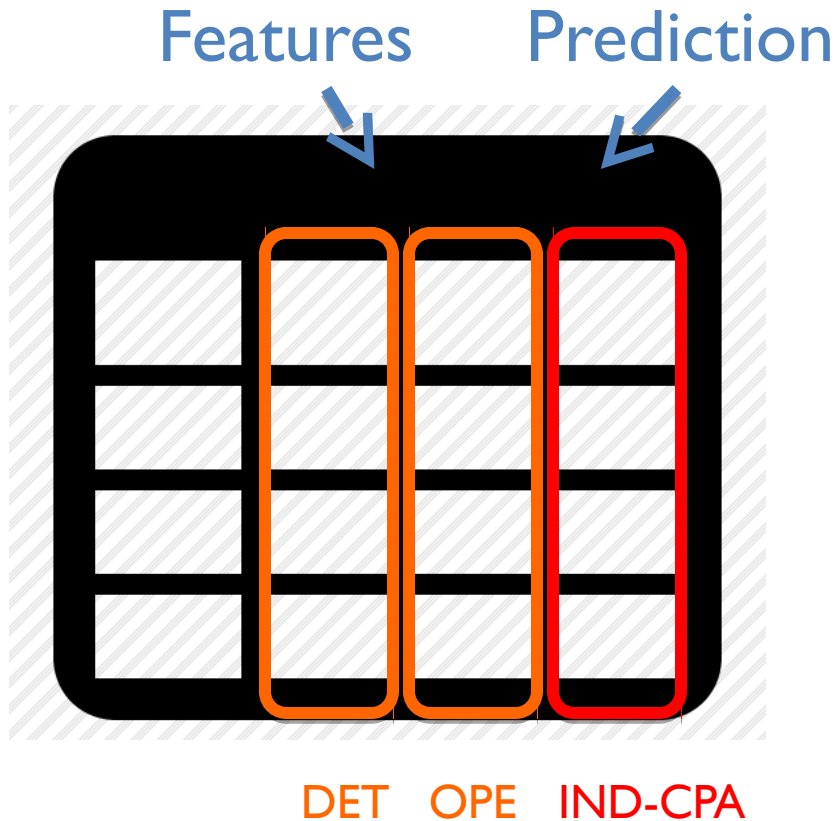
Density of
multinomial distribution

$$\Pr\{\vec{c} \mid \mathbf{f} = f; \rho\} = \Pr\{c_1, c_2, \dots, c_n \mid \mathbf{f} = f; \rho\} = K_c \prod_{i=1}^m \rho_i^{c_{f(i)}}$$

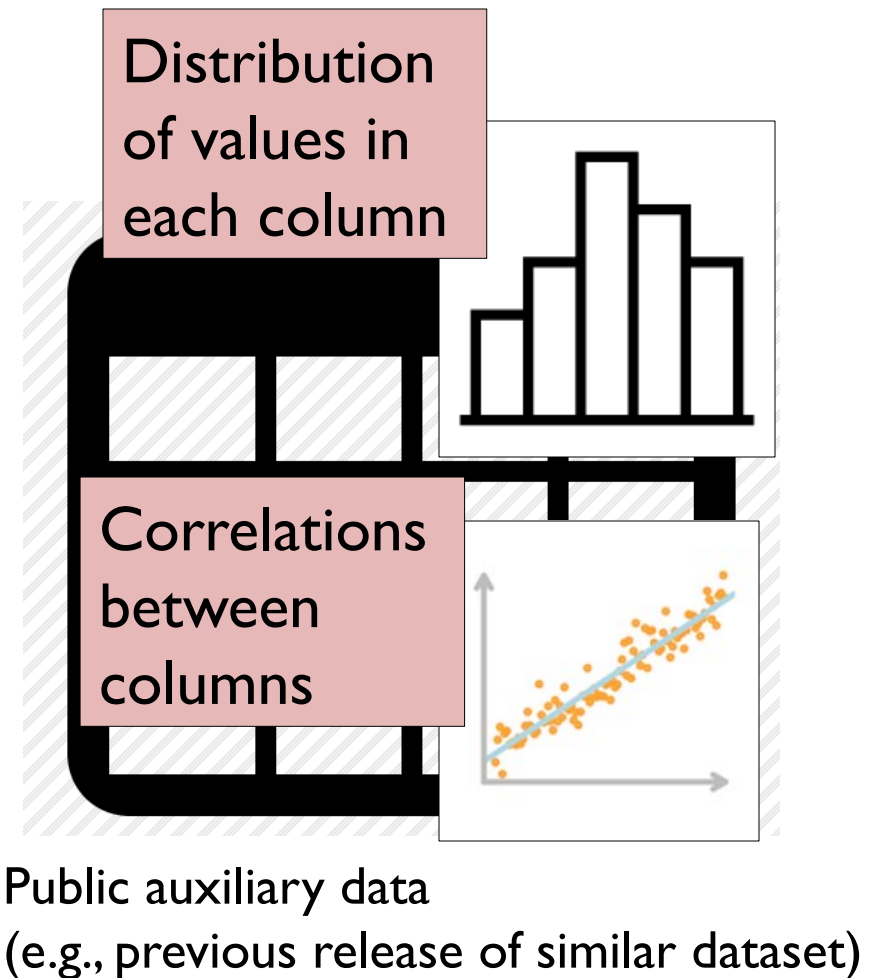
Multinomial Attack

- Optimal
 - Maximum likelihood estimator for deterministic ORE
- Outperforms previous heuristics
 - Naveed et al. frequency analysis (CCS 2015)
 - Grubbs et al. non-crossing attacks (Oakland 2017)
- Extends to multiple columns
 - Condition distribution on previously recovered plaintexts for a dependent column

Inferring “Sensitive” Columns



Multinomial attack!



Let's Try with Real Data



- Over 7 million hospital discharge records each year
- Demographic + medical attributes



- Over 3 million records each year
- Demographic attributes, income

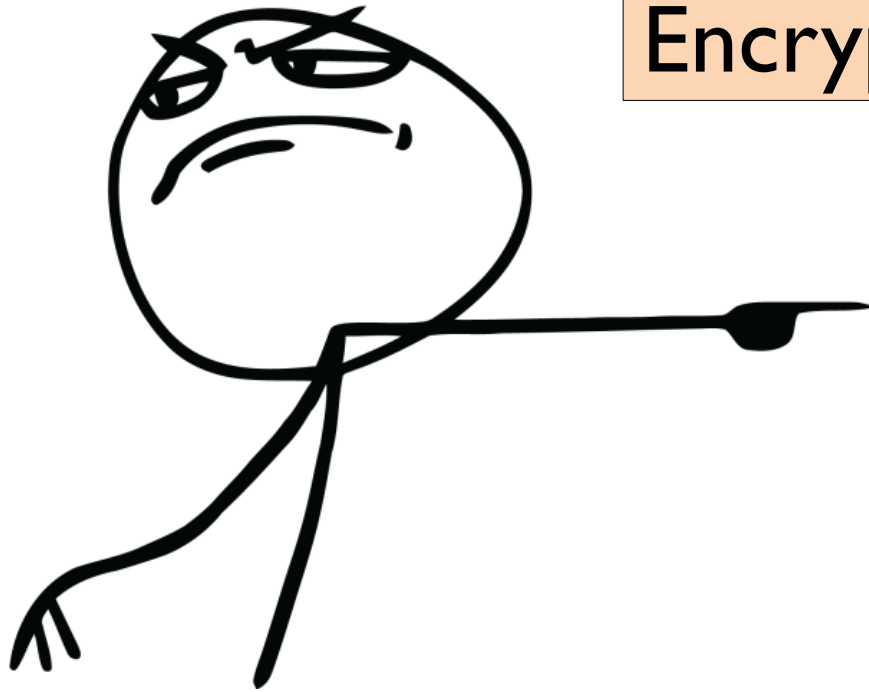


- Data dump from 2015 hack
- Names and addresses of over 600,000 police officers

Empirical Results

- HCUP-NIS hospital discharge records
 - Infer if patient has a mental health or substance abuse condition with 97% accuracy
 - ... mood disorder with 96% accuracy
- U.S. Census American Community survey
 - Recover 90% of PRE-encrypted attributes
 - Infer income to within \$8.4K
- Fraternal Order of Police (FOP) data dump
 - Exact home addresses of 5,500 police officers in PA

Remember



Encryption scheme is “secure”

does not mean

The system is “secure”

Advice to Practitioners

