

NSAC

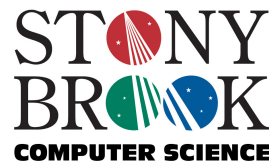
Network Security and Applied
Cryptography Laboratory

<http://crypto.cs.stonybrook.edu>

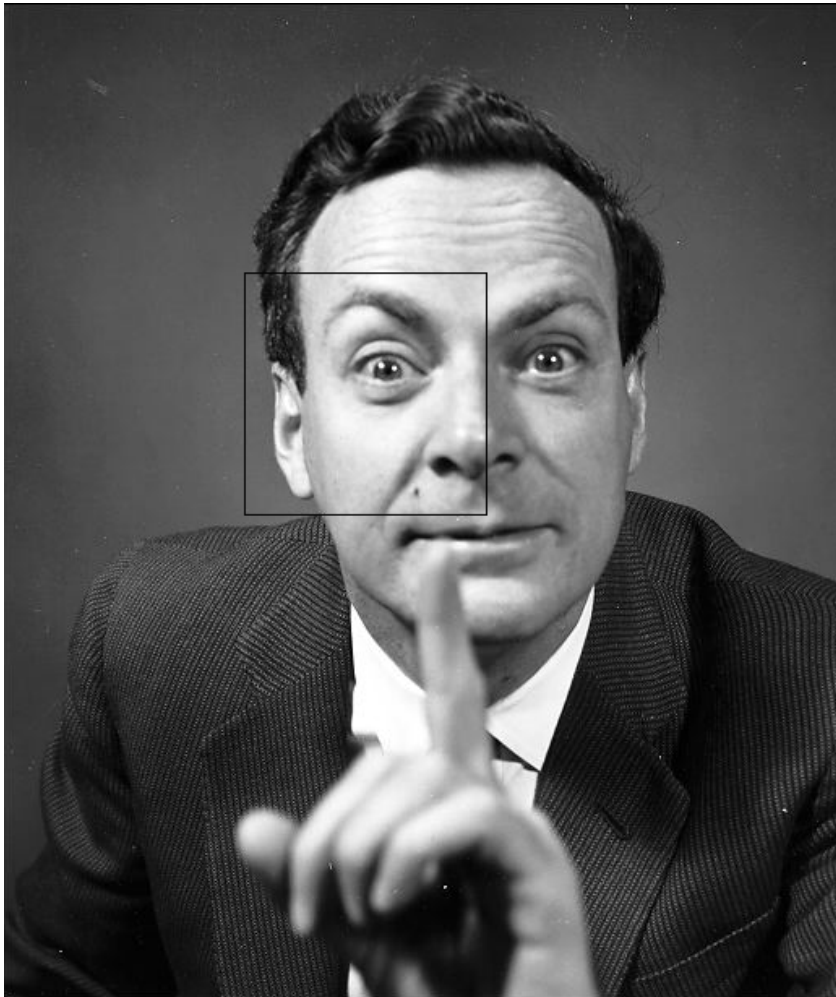
Secure Data Outsourcing

Tutorial @ COMAD 2006, New Delhi, India

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Feynman moment



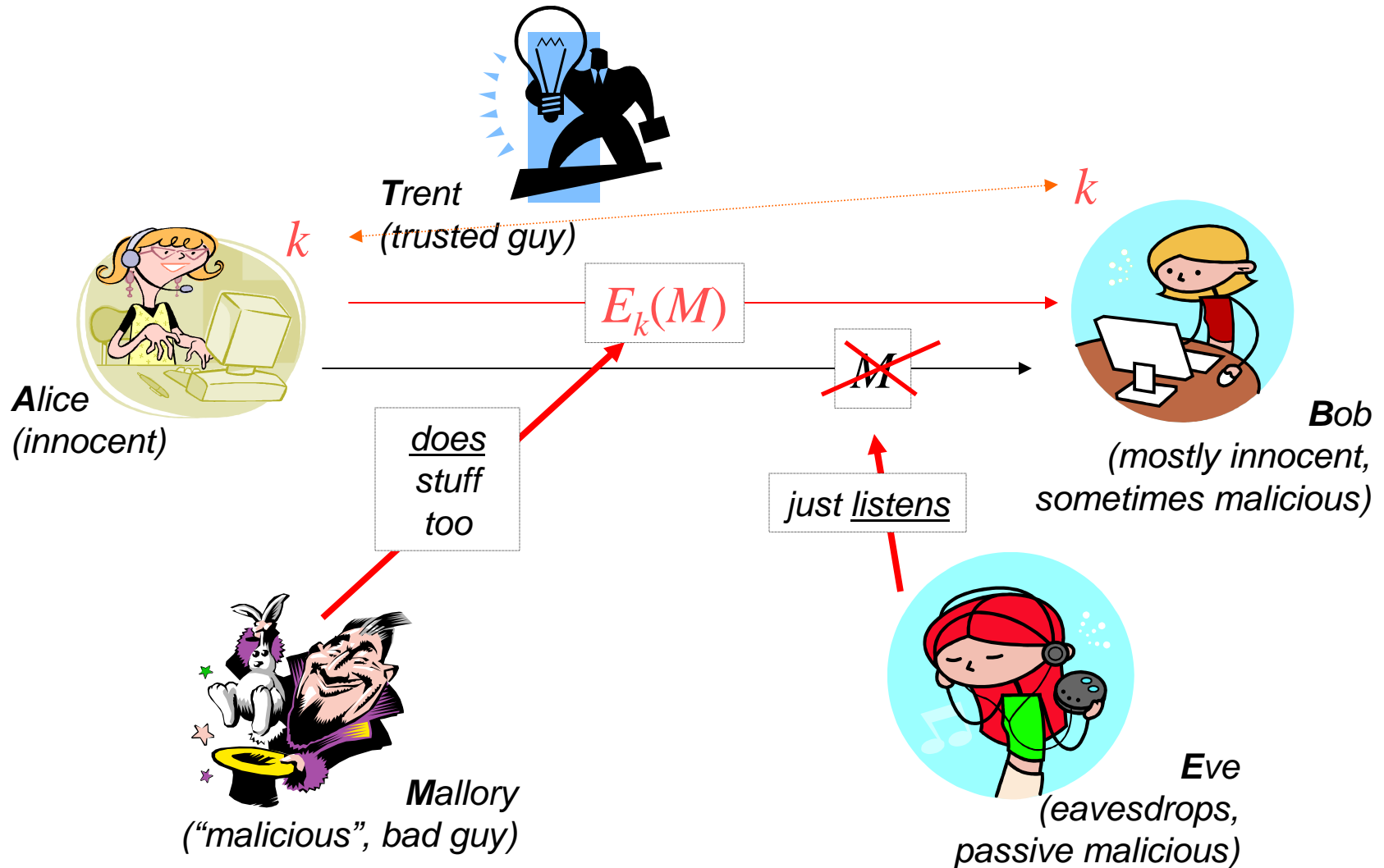
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"I have much experience only in teaching graduate students [...] and as a result [...] I know that I don't know how to teach."

- ➔ *Crypto Crash Course*
- ➔ *Data Outsourcing*
- ➔ *Query Correctness*
- ➔ *Data Confidentiality*
- ➔ *Access Privacy*
- ➔ *Searching on Encrypted Data*
- ➔ *Trusted Hardware*

- *Randomness*
- *Crypto Hashes*
- *Encryption*
- *Public key encryption*
- *Signatures*
- *Ciphers*
- *Semantic Security*
- *Forward Secrecy*
- *Performance*
- *Merkle/Hash trees*

Crypto: Meet the cast



Cryptographically random numbers: a sequence of numbers X_1, X_2, \dots such that for any integer $k > 0$, it is **impossible** for an observer to predict X_k even if all of X_1, \dots, X_{k-1} are known.

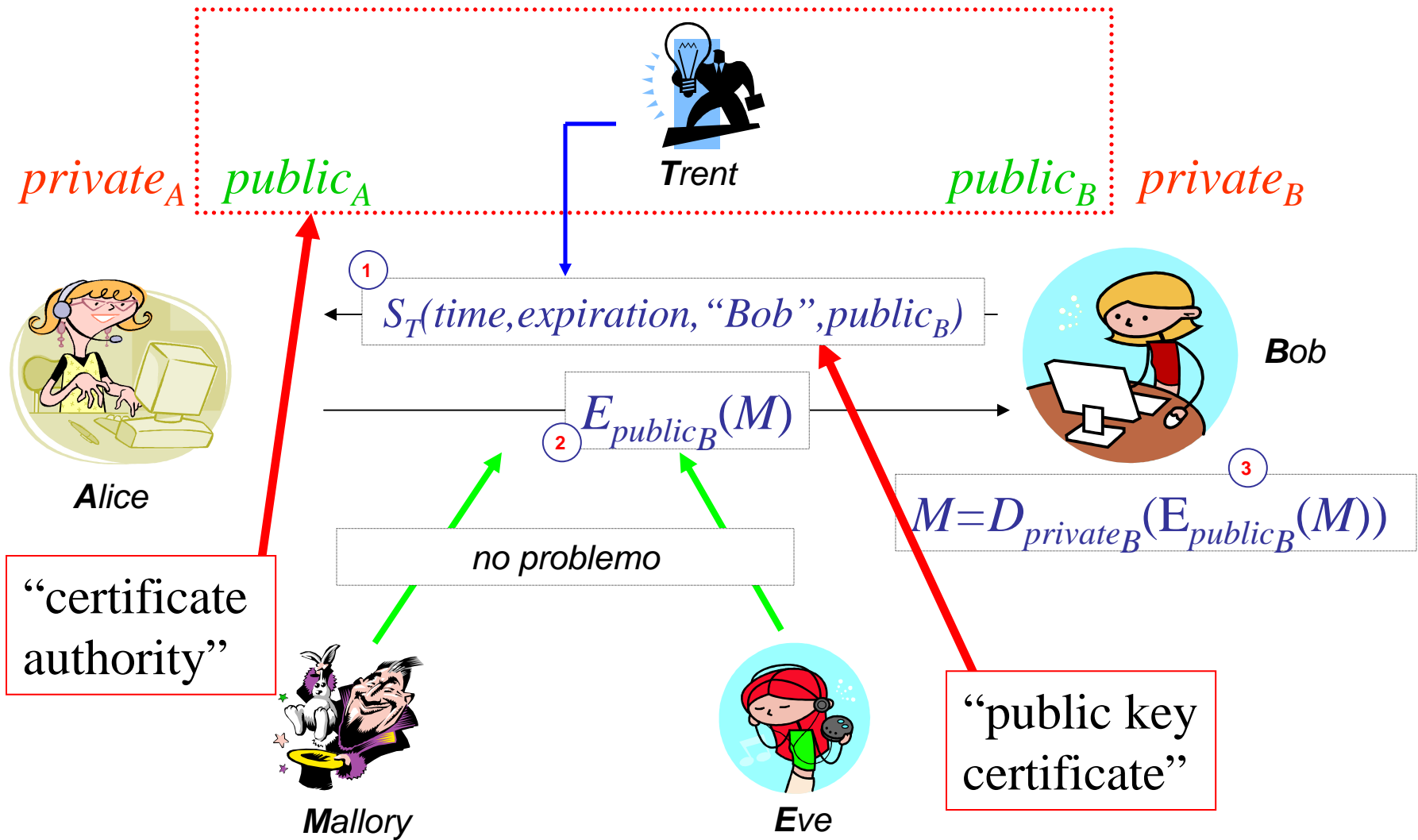
Problem: True RNGs cannot be deterministically algorithmic in a closed system. “Anyone who considers arithmetic methods ... is in a state of sin” (von Neuman)

Being creative: simulate a sequence of cryptographically random numbers but generate them by an algorithm.

Pseudo-random numbers: a sequence of numbers X_1, X_2, \dots such that for any integer $k > 0$, it is **hard** for an observer to predict X_k even if all of X_1, \dots, X_{k-1} are known.

- A hash is a **one-way, non-invertible** function of that produces **unique** (with *high likelihood*), **fixed-size** outputs for different inputs.
- The probability of any bit “flipping” in the output bit-string should be always $\frac{1}{2}$ for any change (even one bit) in the input (“randomness”).

Crypto: PKI

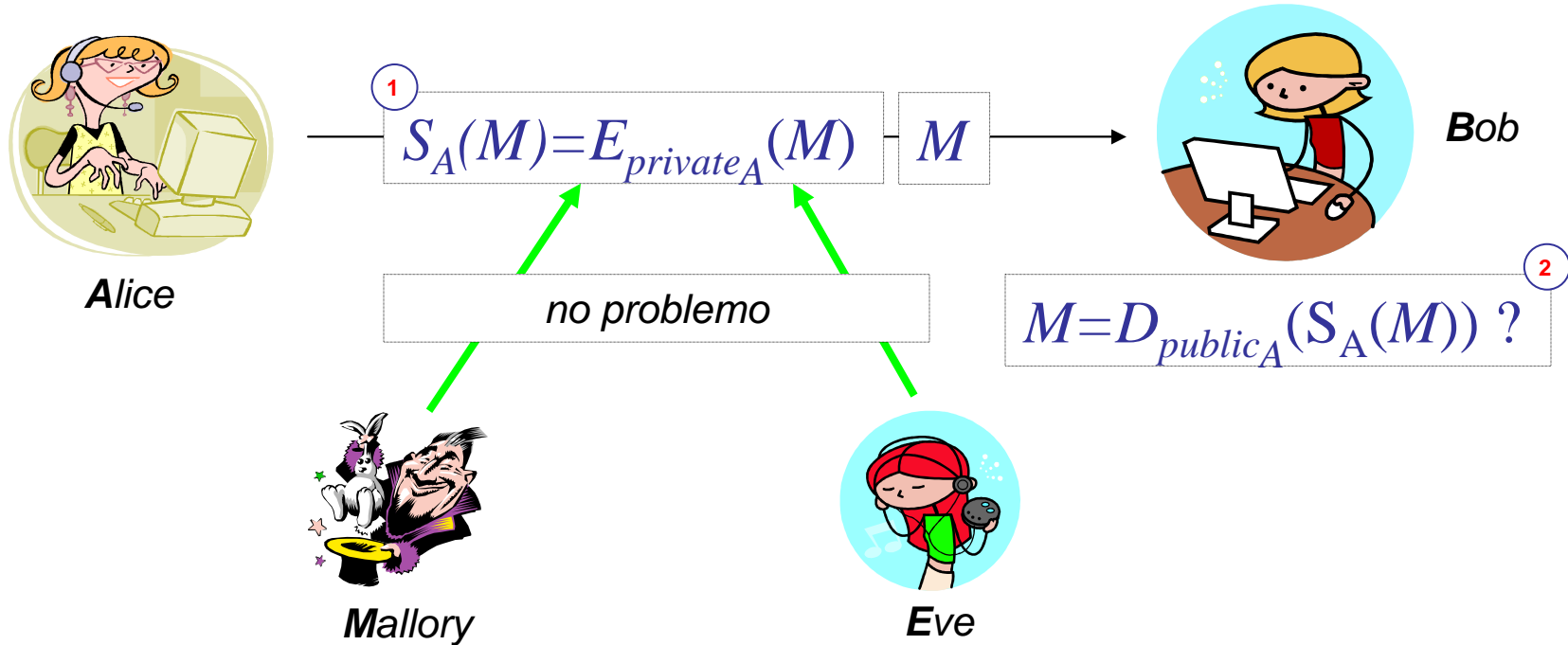


Crypto: Signatures

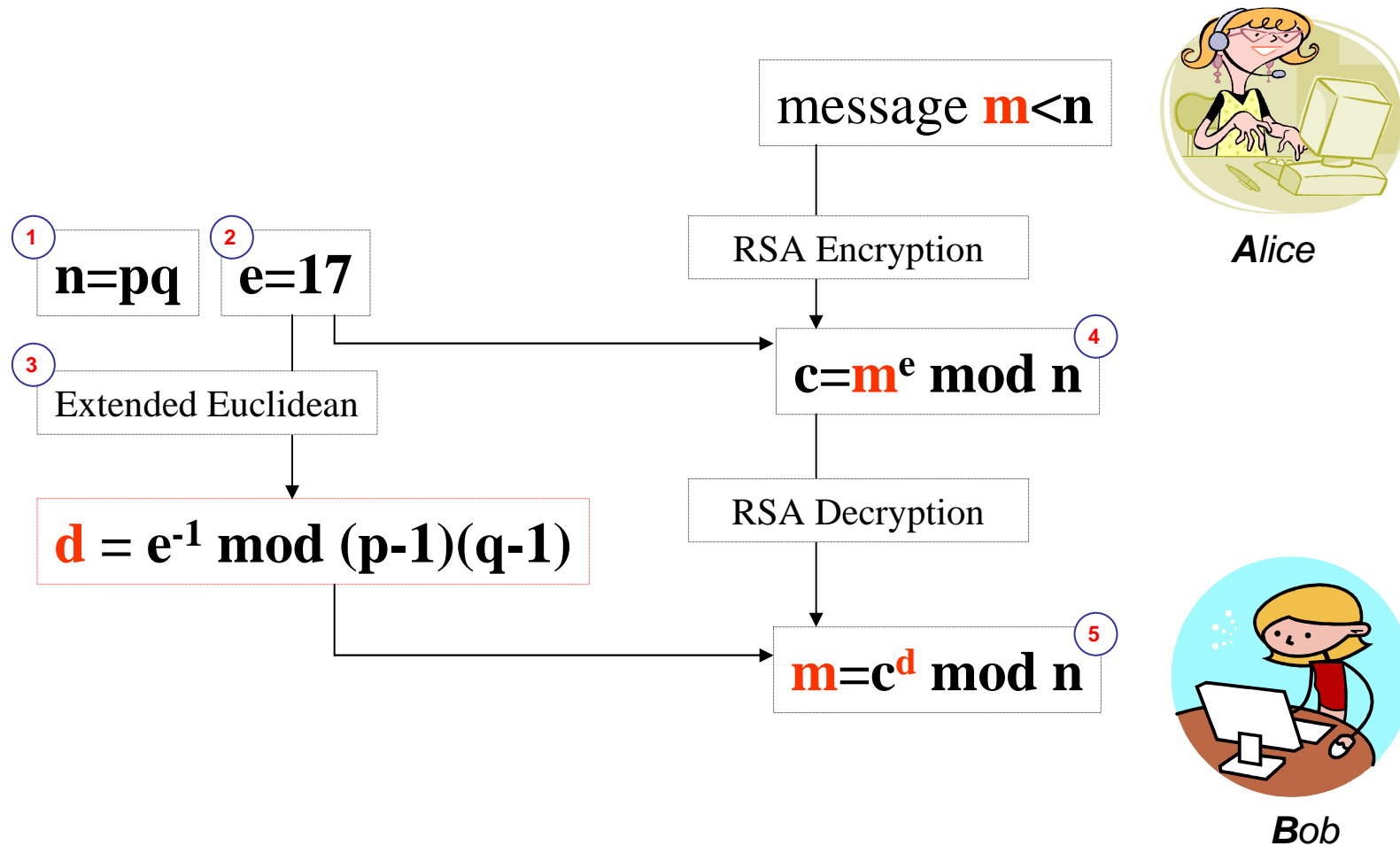
$$M = D_{private_A}(E_{public_A}(M)) = D_{public_A}(E_{private_A}(M))$$

private_A *public_A*

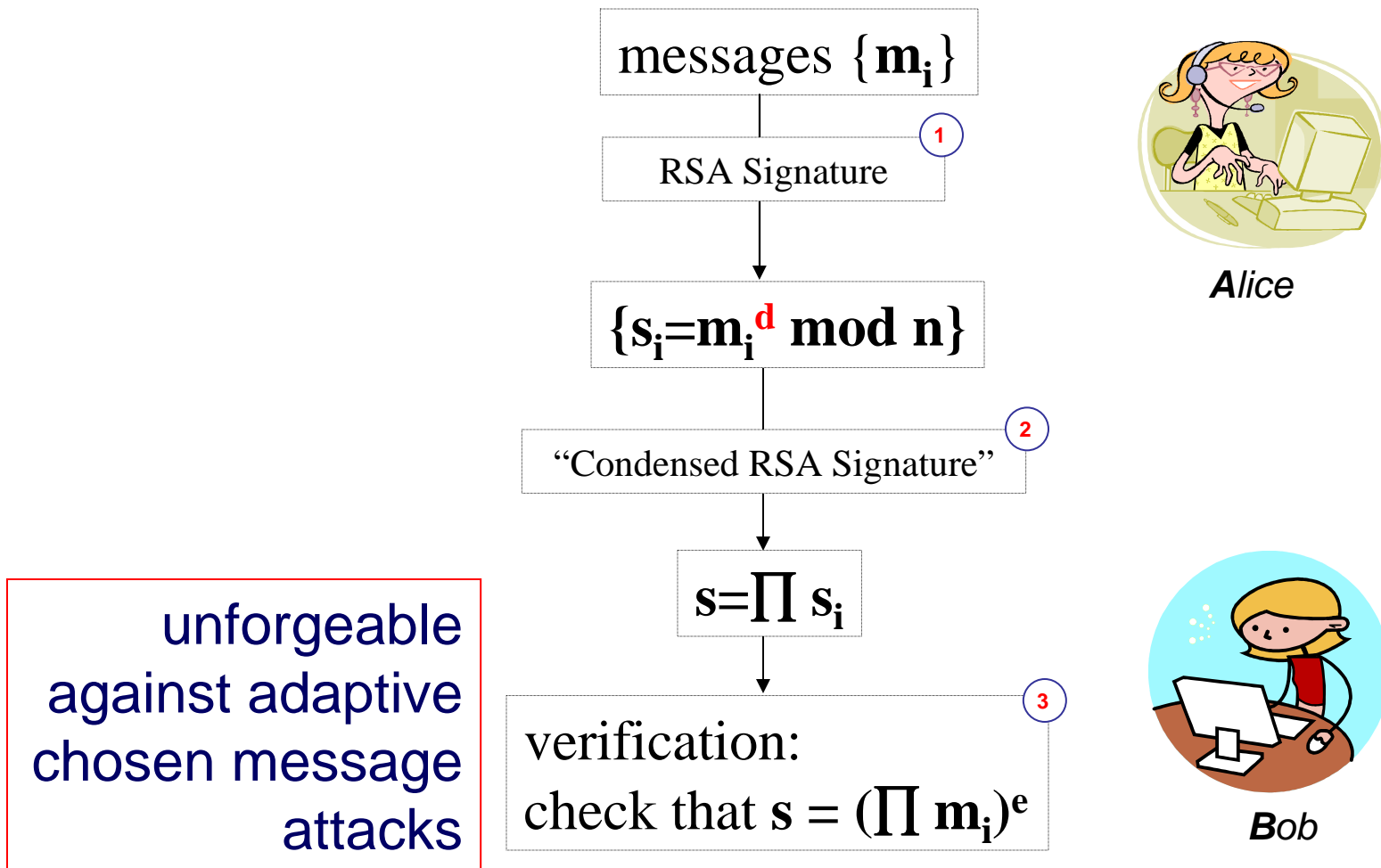
public_B *private_B*



Crypto: RSA in a nutshell

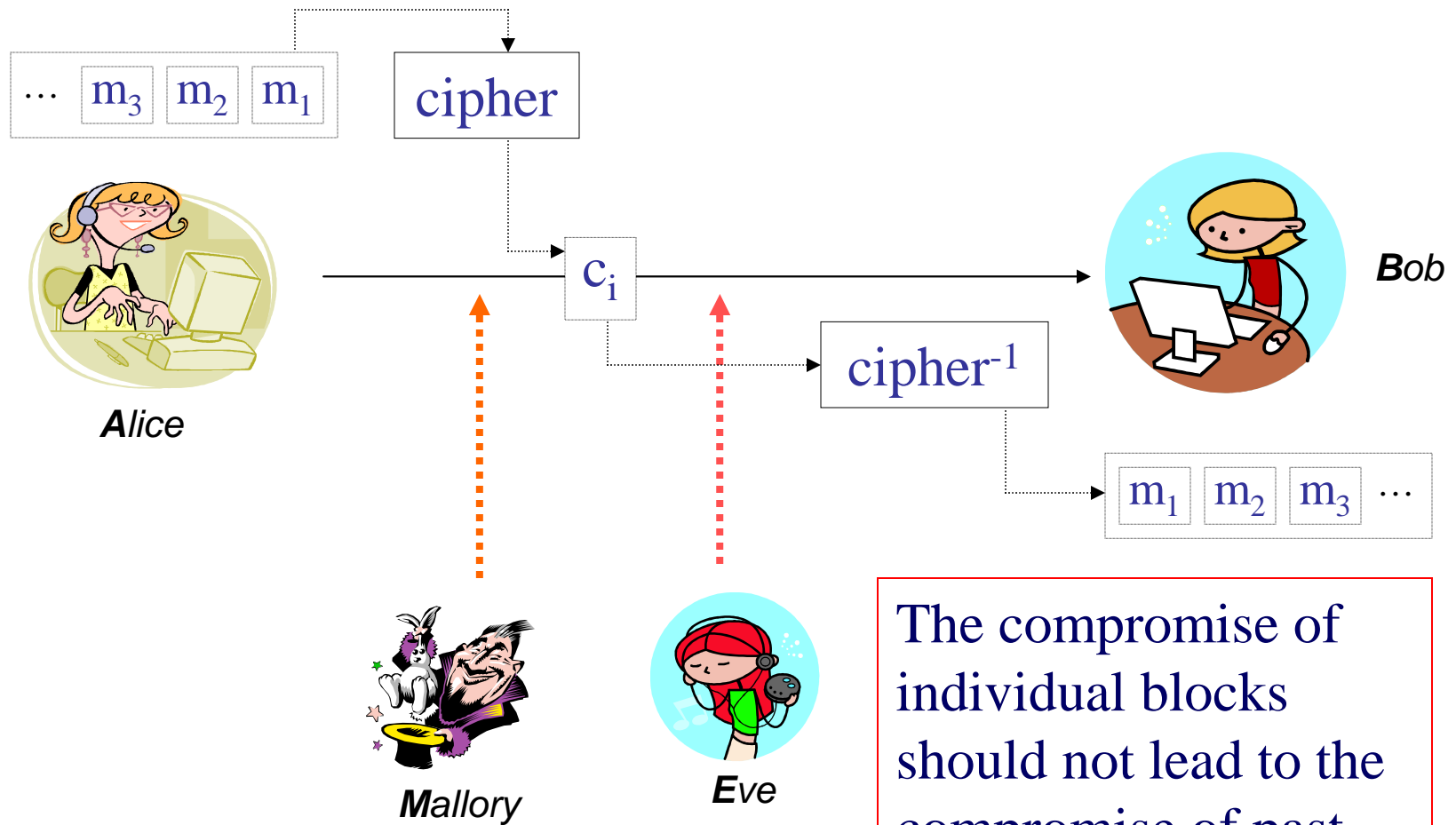


Crypto: Condensed RSA

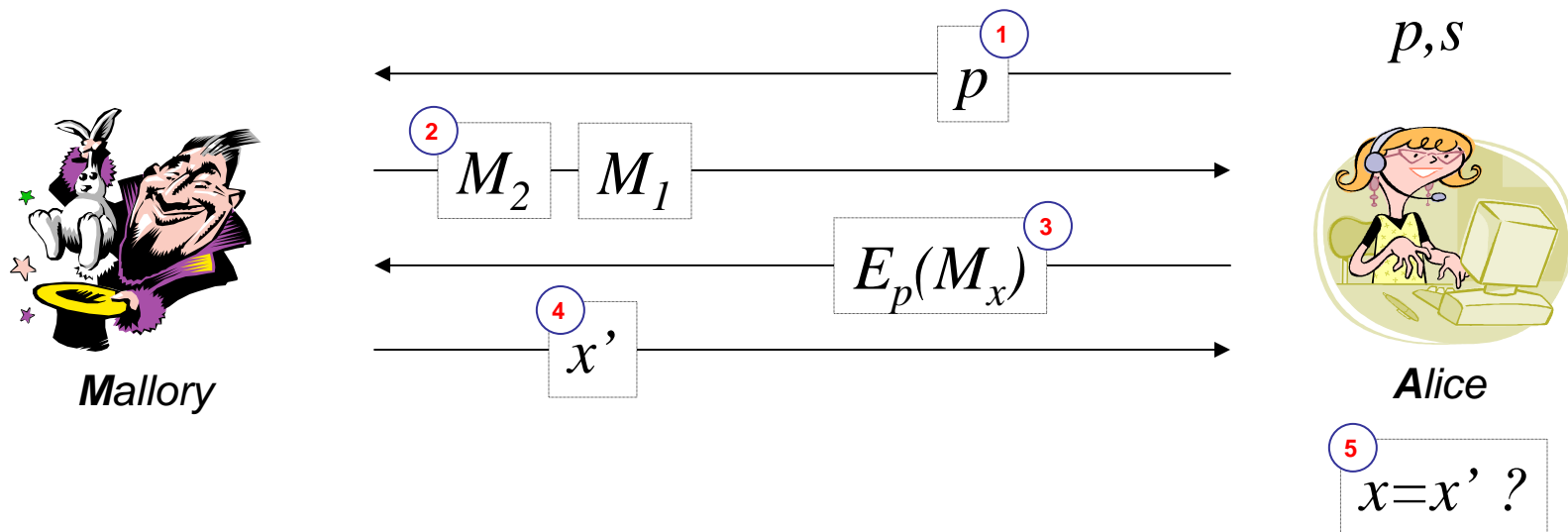


unforgeable
against adaptive
chosen message
attacks

Crypto: Ciphers

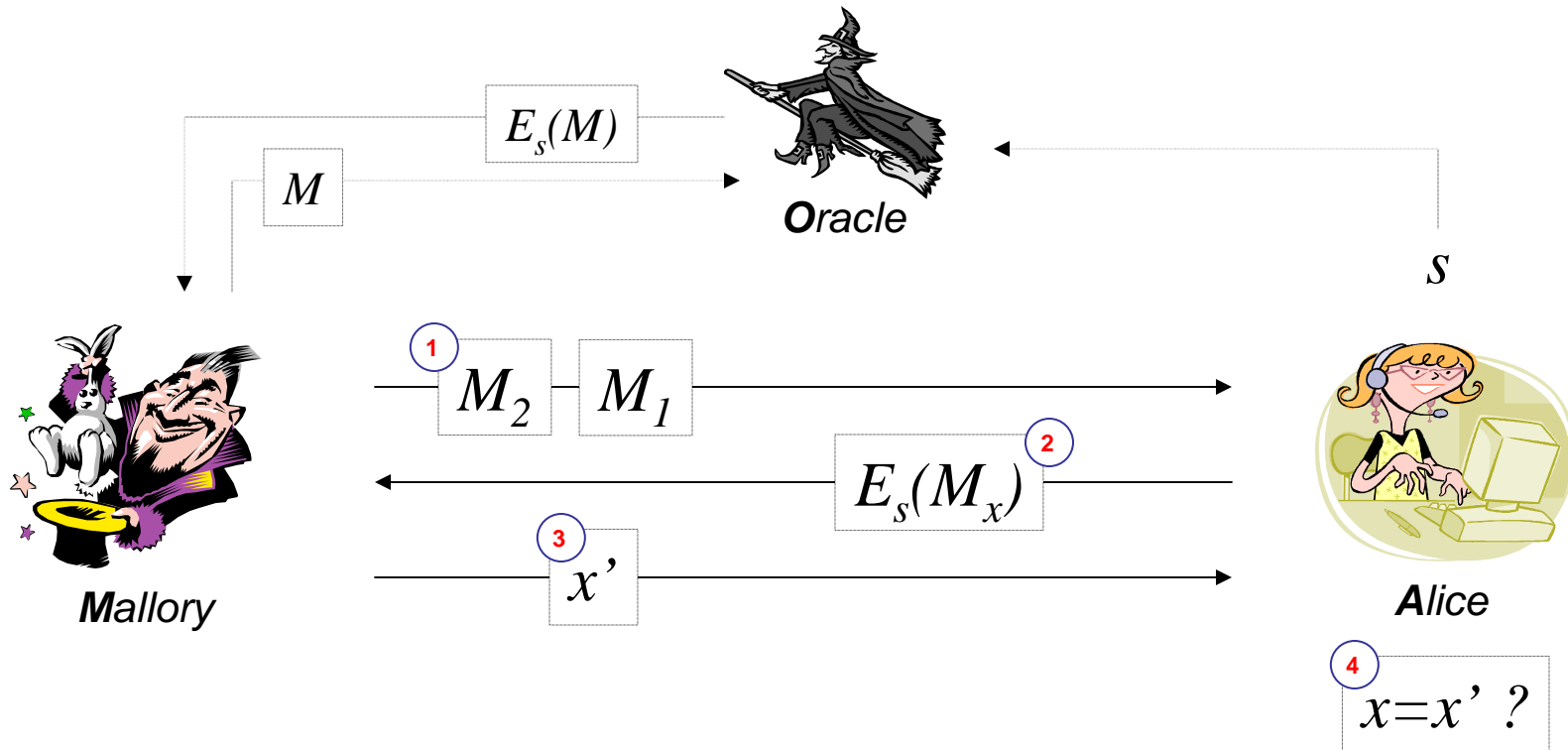


$\text{len}(M_1) = \text{len}(M_2) ?$



$E()$ is **indistinguishable under a chosen plaintext attack** (IND-CPA, “semantically secure”) if no probabilistic polynomial time-bounded Mallory can succeed in finding x' , significantly better than guessing.

Crypto: Semantic Security



- Deterministic + stateless = insecure !
- Semantic security implies *bit security* !
- RSA : non-semantically secure ! Why ?!
- RSA + padding (e.g., RSA-OAEP): ok

Future compromise (e.g., of PK secrets) should not propagate backwards in time.

Crypto: Performance



Illustrative baseline.

Pentium 4. 3.6GHz.
1GB RAM. 11000 MIPS.
OpenSSL 0.9.7f

DES/CBC: **70MB/sec**

RC4: **138MB/sec**

MD5: **18-615MB/sec**

SHA1: **18-340MB/sec**

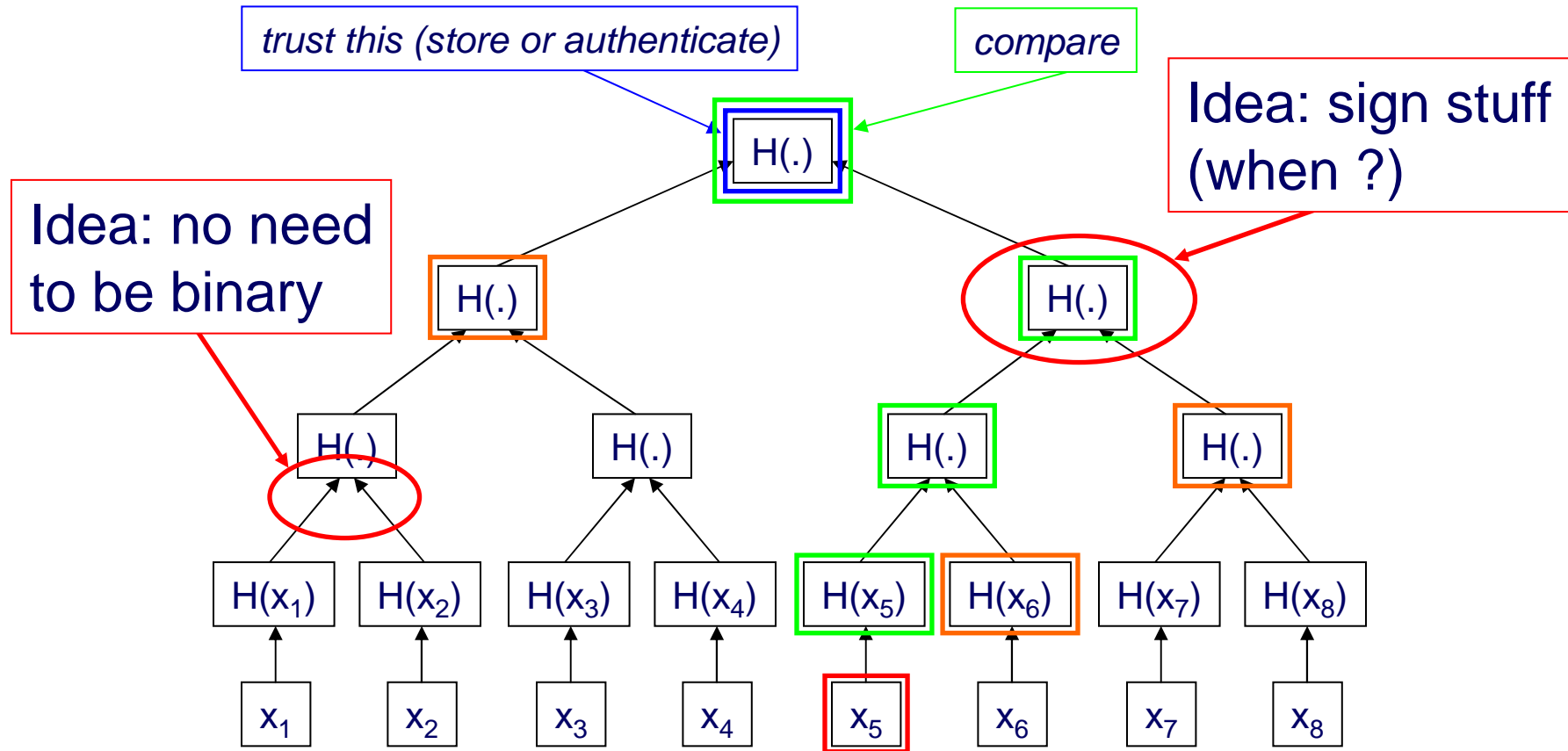
Modular MUL 1024: **273000/sec**

RSA1024 Sign: **261/sec**

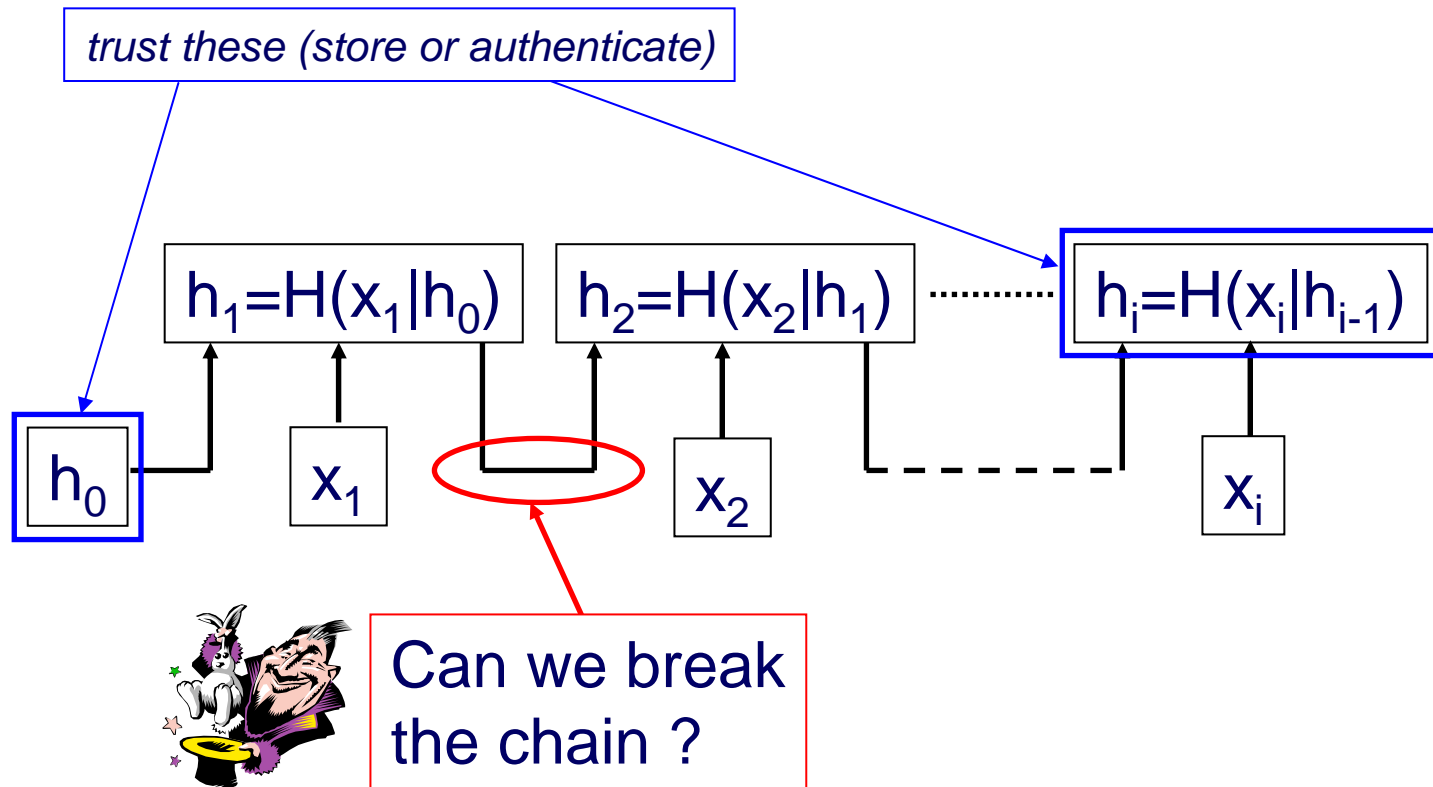
RSA1024 Verify: **5324/sec**

3DES: **26MB/sec**

Crypto: Merkle/Hash trees



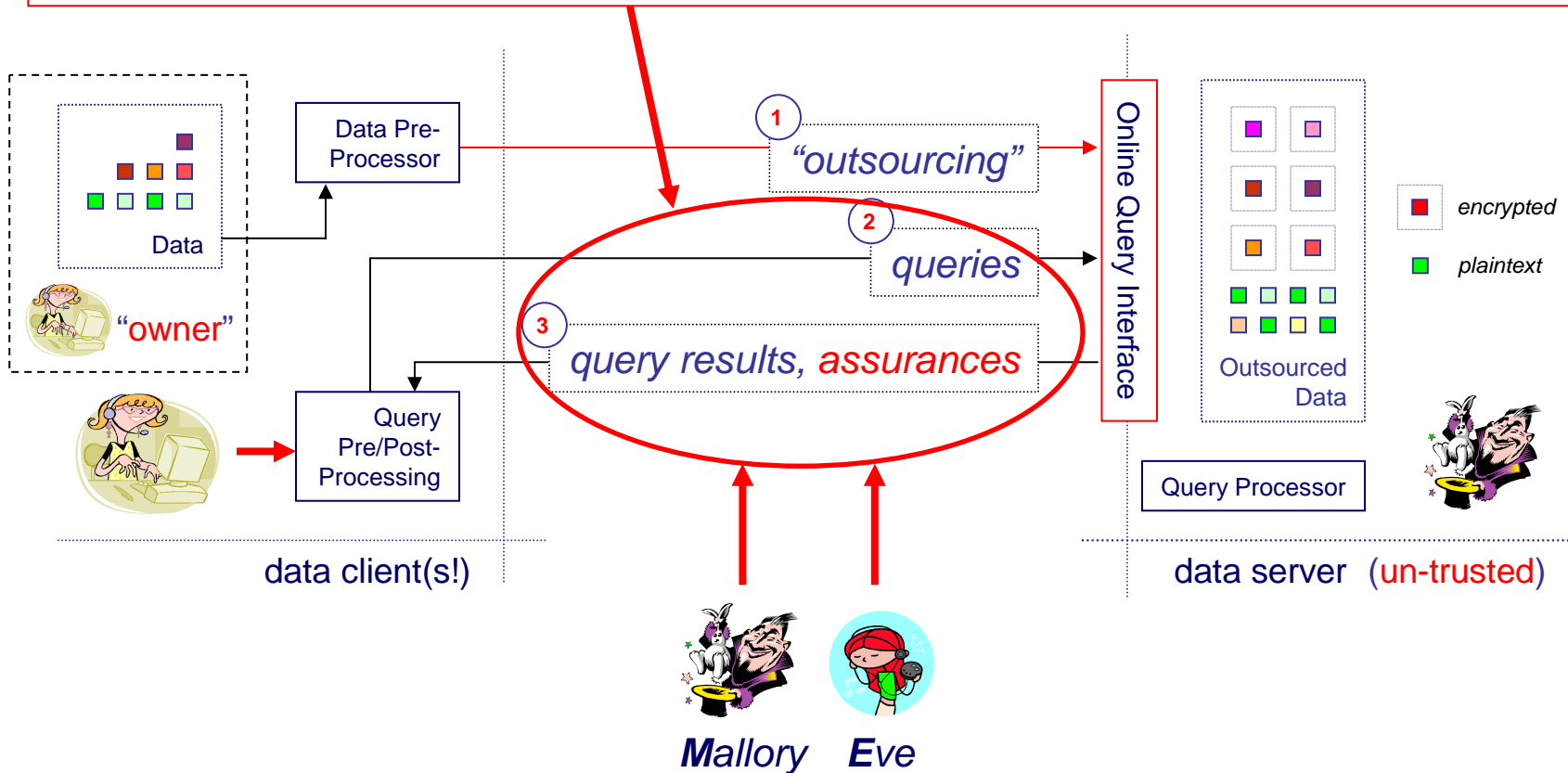
Crypto: Hash chains



- ➔ *Crypto Crash Course*
- ➔ *Data Outsourcing*
- ➔ *Query Correctness*
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- ➔ *Trusted Hardware*

Data Outsourcing

$assurances \subseteq \{query\ correctnes, data\ confidentiality, access\ privacy\}$



Outsourcing Challenges

Un-trusted server:

- lazy: incentives to perform less
- curious: incentives to acquire information
- malicious:
 - denial of service
 - incorrect results
 - possibly compromised

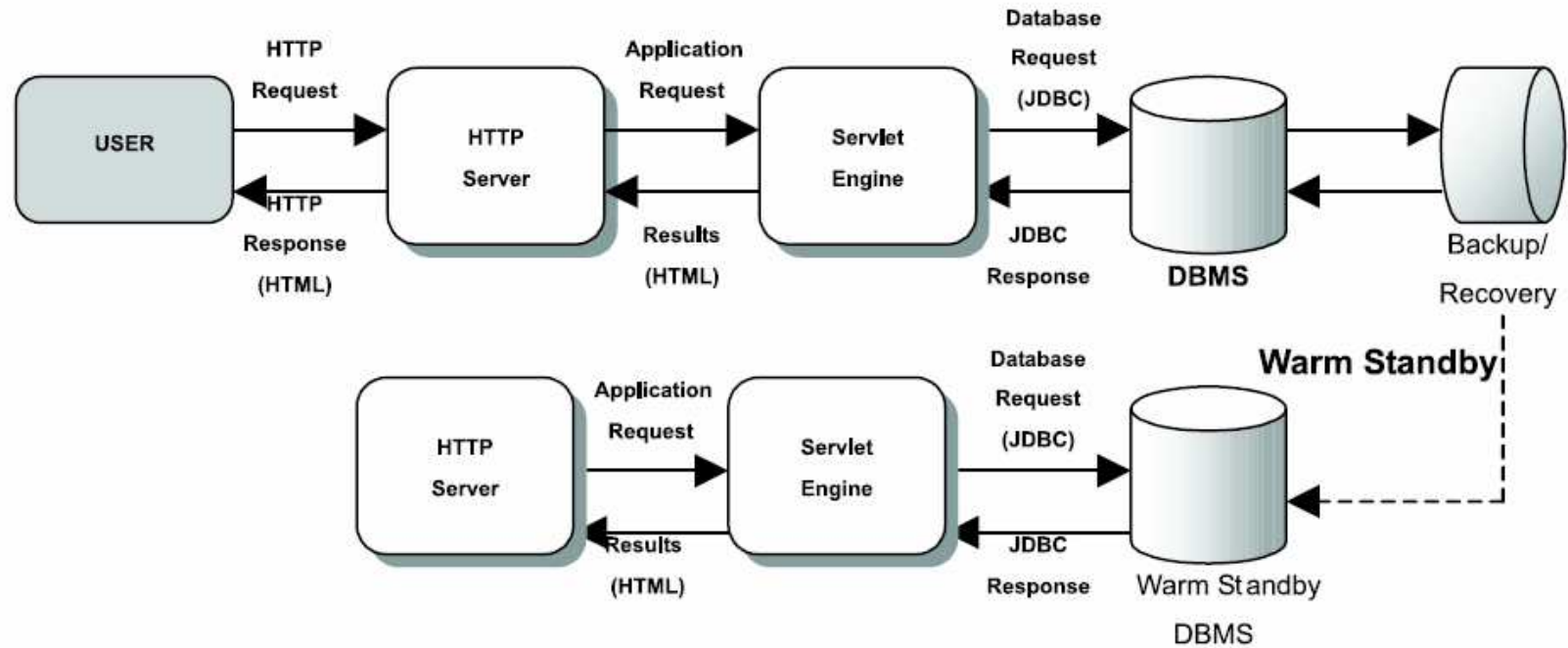
Why is this hard ?

- how ?
- arbitrary expressivity
- overheads
 - network
 - computational costs

What do we do ?

- query assurances
- full privacy
 - of queries (even encrypted)
 - of access patterns
- data confidentiality

Hacigumus (2002)



System architecture of NetDB2

Stored Data Confidentiality

```
SELECT decrypt(discount, key)
FROM lineitem
WHERE custid = 300
```

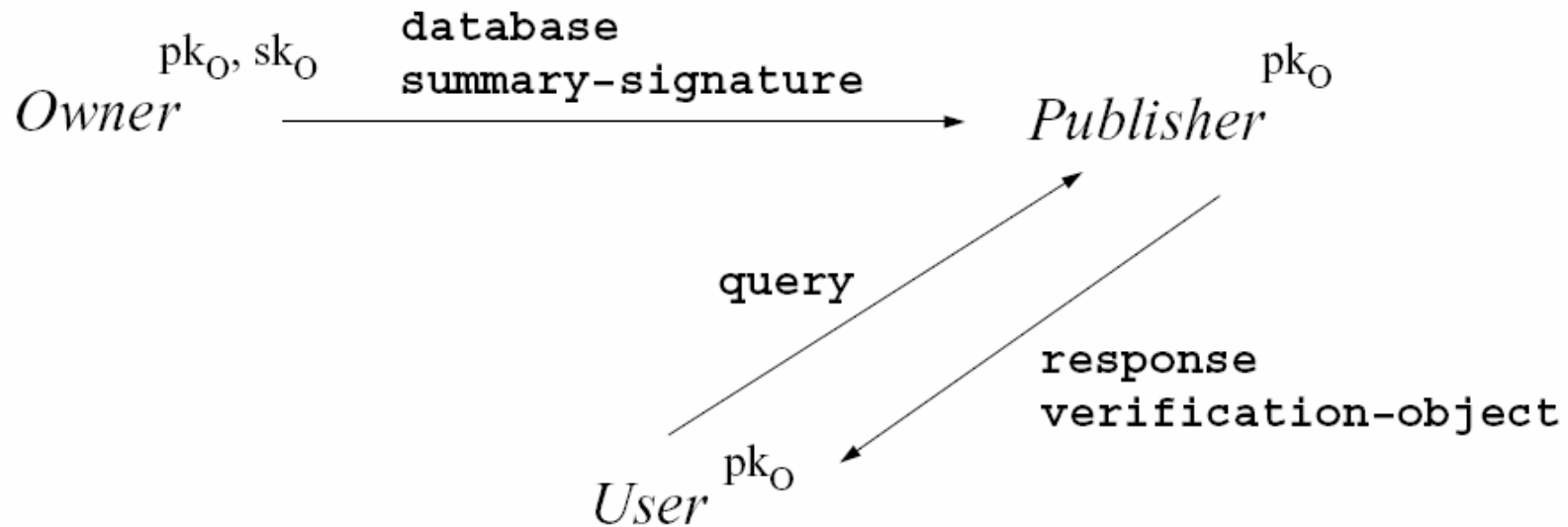
H. Hacigumus, B. R. Iyer, and S. Mehrotra.
Providing database as a service, ICDE 2002.

- ➔ *Crypto Crash Course*
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Client requires quantifiable assurances that query results are correct, for arbitrary query types in the presence of a server that could be ...

... lazy

... and/or fully malicious (!)

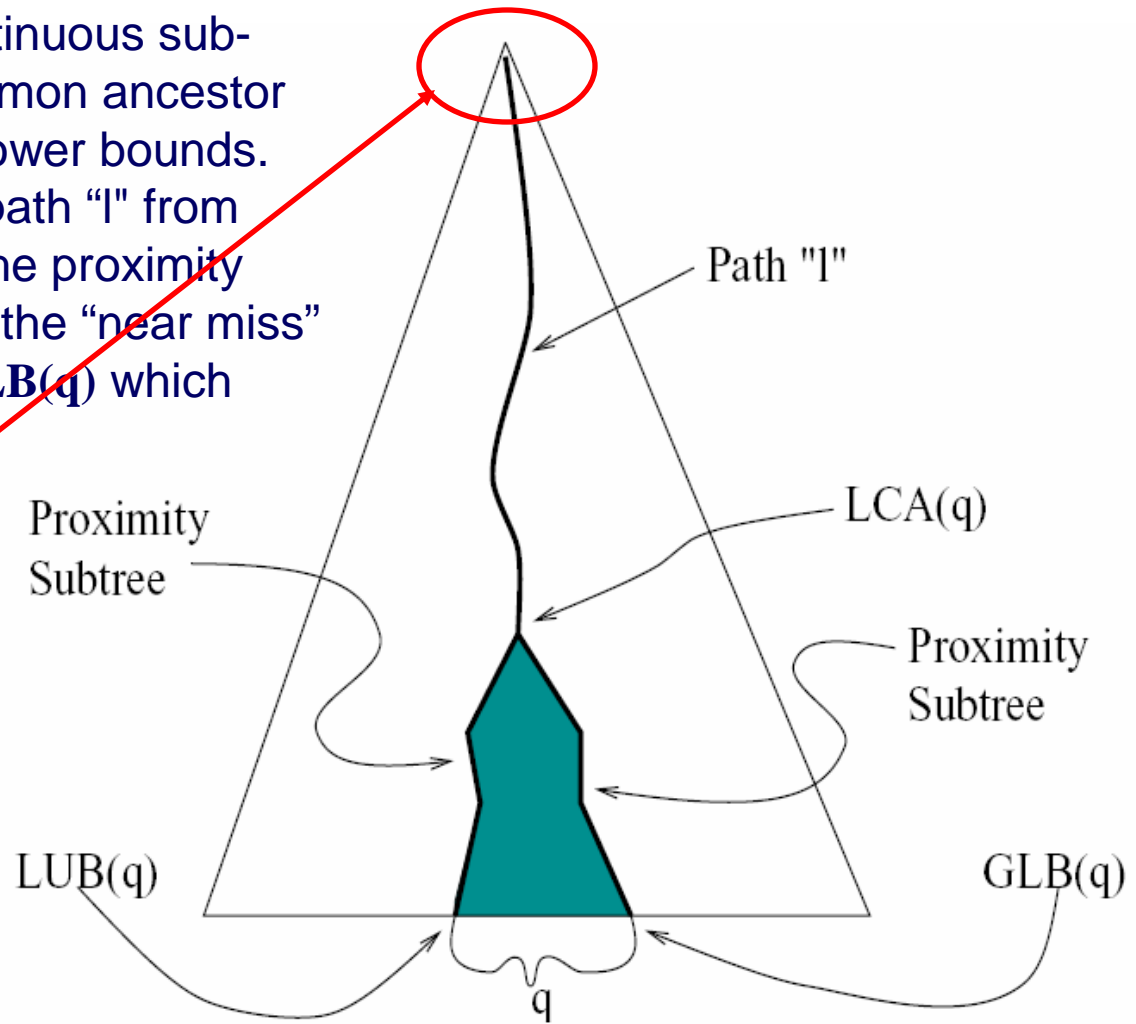


The owner provides database updates and summary signatures to the un-trusted publisher. When users make inquiries with the publisher, they get responses which can be verified using a returned verification-object. Only sk_O is secret, pk_O is authenticated.

Devanbu et. al. (2000)

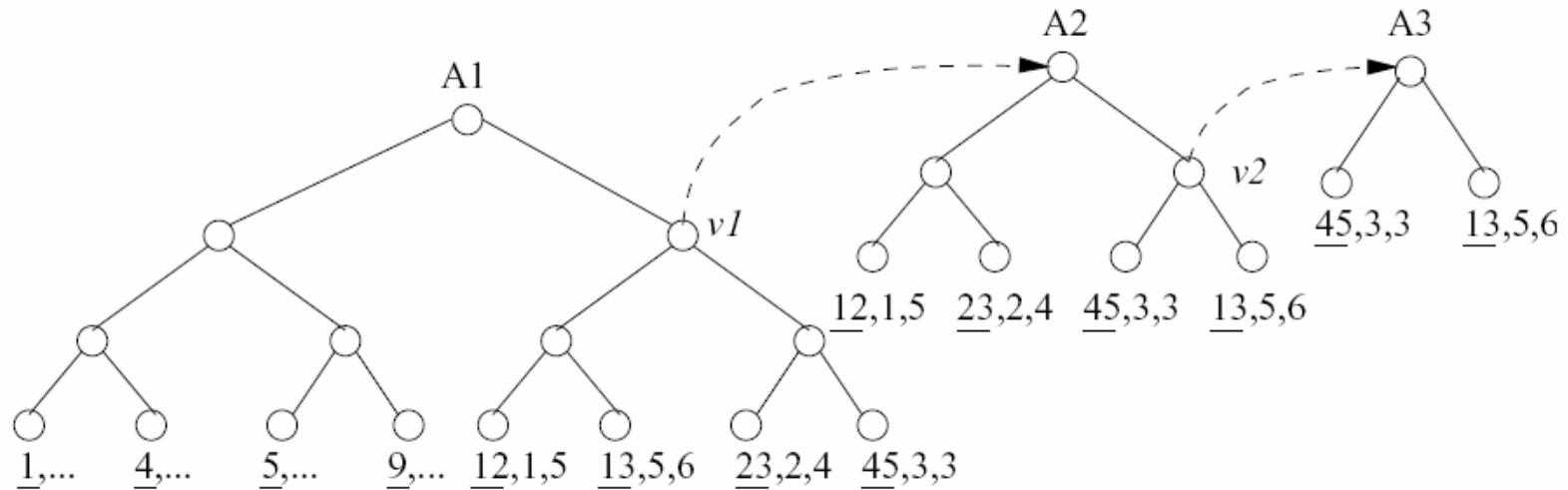
A Merkle tree, with a continuous sub-range q , with a least common ancestor $LCA(q)$, and upper and lower bounds. Note the verifiable hash path "I" from $LCA(q)$ to the root, and the proximity sub-trees (thick lines) for the "near miss" tuples for $LUB(q)$ and $GLB(q)$ which show that q is complete.

authenticated via signature

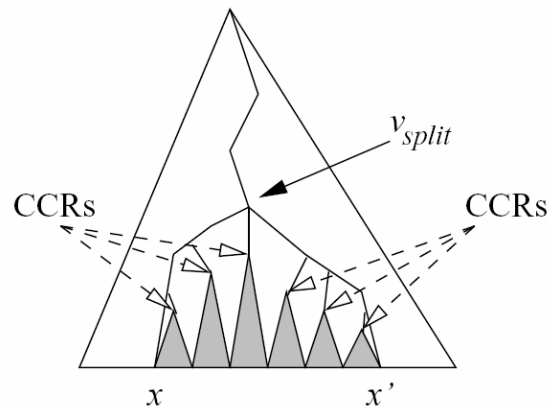


Supported claimed operations:

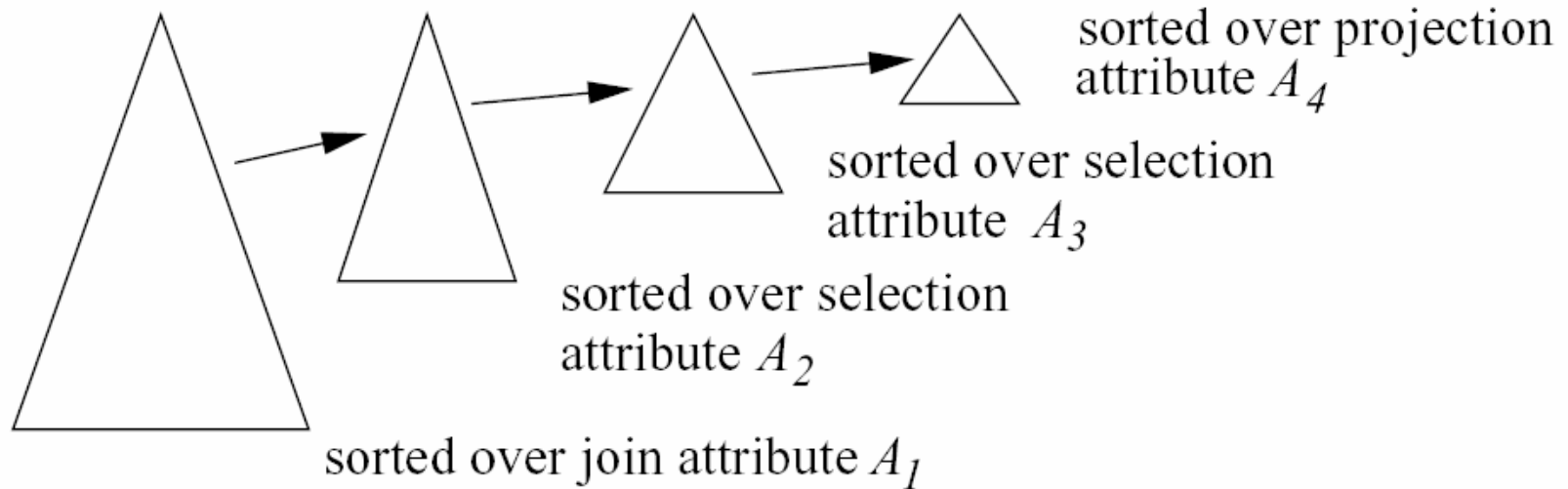
- selections
- projections
 - (1) maintaining VOs before duplicate elimination
 - (2) pre-computing VOs for common projections
- equiJOIN
 - (1) keep materialized cartesian product $S \times R$
 - construct VO on sorted version of product (according to difference (S.A-R.A)) – this yields 3 types of leaf nodes (“0”, “<”, “>”) in Merkle tree
 - (2) all kinds of other tricks
- set operations
 - union (client does it and verifies VOs for input sets)
 - intersection (?)
 - multi-dimensional range queries (generalizing hash tree to “multi-dimensional range tree”)



Excerpt of a 3-dimensional range tree, sorted by attributes A_1, A_2 and A_3



Covering canonical roots (CCR):
 roots of the canonical sub-trees
 precisely covering the leaves
 with values in the interval.



```
SELECT S.A4 FROM S,R  
WHERE S.A1=R.A1 AND A2<10 AND A3>17
```

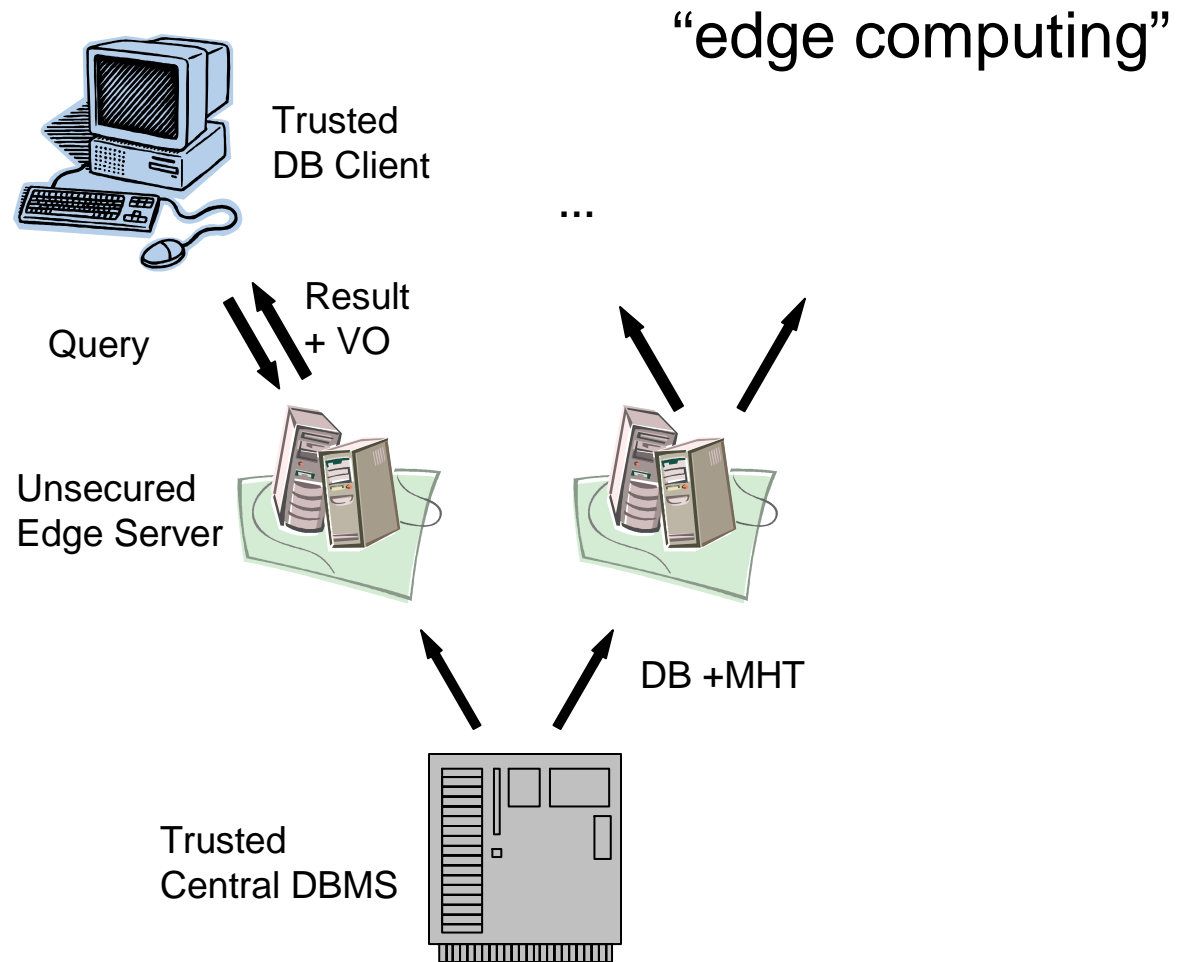
Issues:

- query expressiveness
- query flexibility
 - works only on data with VOs
- “universe split” phenomenon
 - use timestamps, expiration times
- expensive operations (!)

Discusses the use of batch verification of signatures and similar techniques (condensed RSA) to authenticate results.

		Condensed-RSA	Batch-DSA	BGLS
Sign	1 signature	6.82	3.82	3.54
Verify	1 signature	0.16	8.52	62
	t = 1000, k = 1	44.12	1623.59	184.88
	t = 100, k = 10	45.16	1655.86	463.88
	t = 1000, k = 10	441.1	16203.5	1570.8

Cost comparison (in msec): verification and signing. Notation: t – # signatures, k – # signers



Claimed problems with [Devanbu 2000]

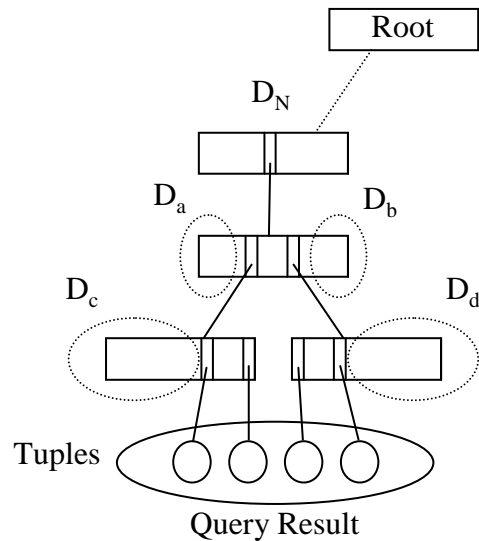
- A hash tree is needed for every sort-order
- VOs need to contain links all the way to the root,
 - VOs grow linearly to query result and logarithmic to base table size
- Projections may have to be performed by clients
- No provision for dynamic updates on the database

Aim 1: VO size just linear in query result

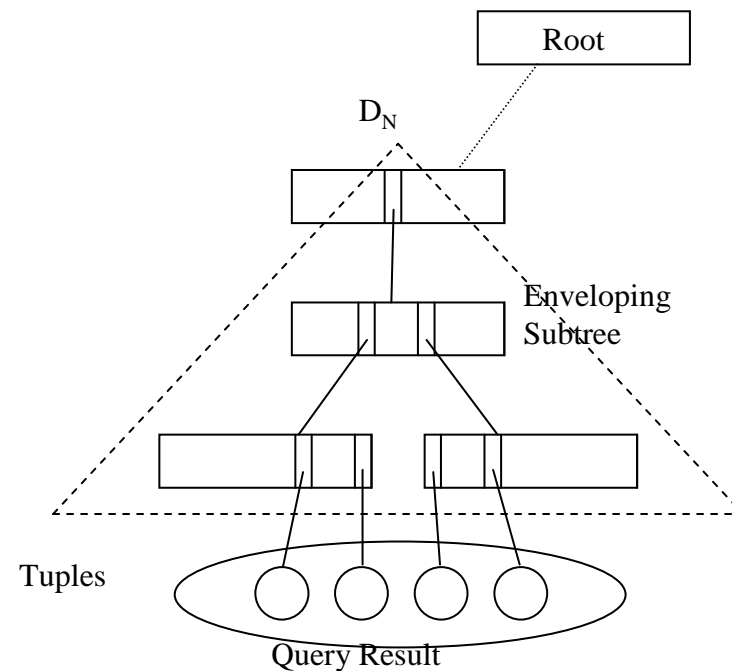
Aim 2: do not push projections to client

Idea: use different hash function

- $h(x) = g^x \text{ mod } q$
- h is commutative, $h(x+y) = h(y+x)$
 - Digests can be combined arbitrarily
 - Projection can be performed at the edge servers
 - Facilitates insertion of new tuples with minimal effect on other digests
- but: significantly (1000-10000 times) slower
- trade-off: computation vs. communication



Verification object = $D_N + D_S$
 where $D_S = \{D_a, D_b, D_c, D_d\}$



Verifying Selection
 (no need to go up to the root
 as everything is also signed)

Similar expressiveness. But ...

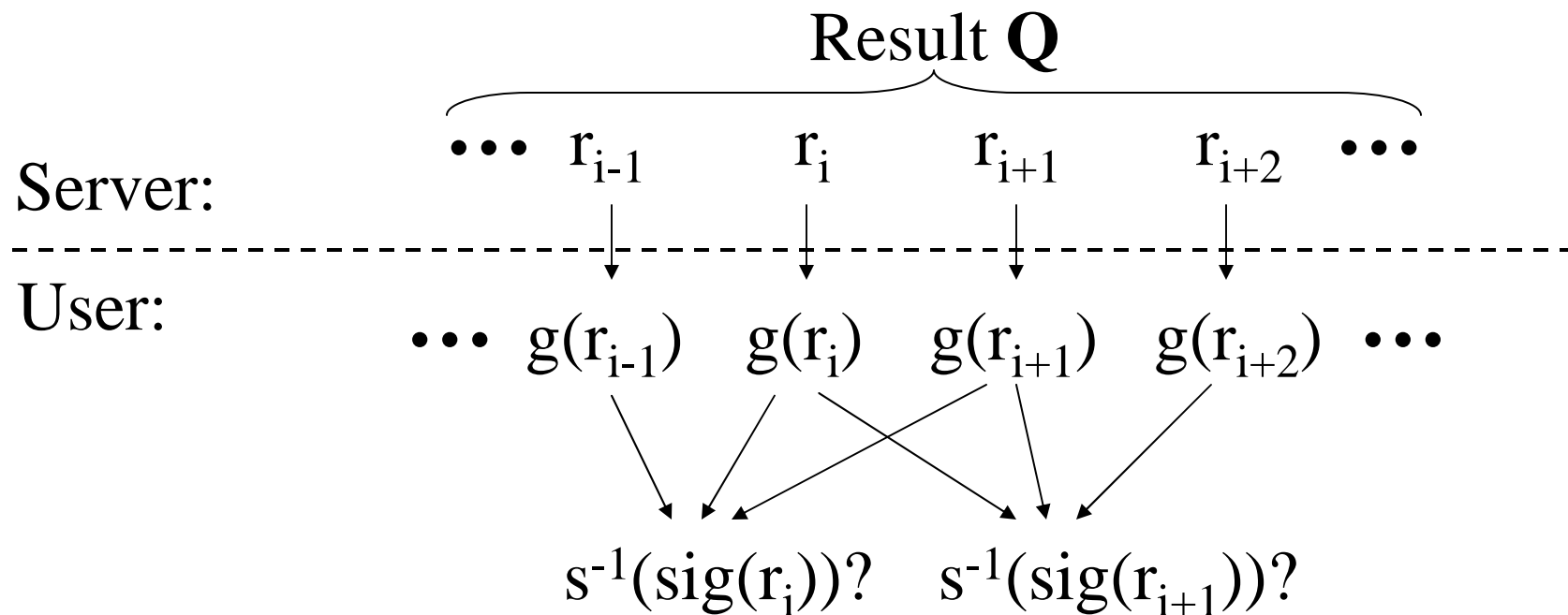
Asks: what about access control rules ?
(Devanbu seems to reveal too much: boundary tuples)

Also claims: lower overheads for queries
and updates.

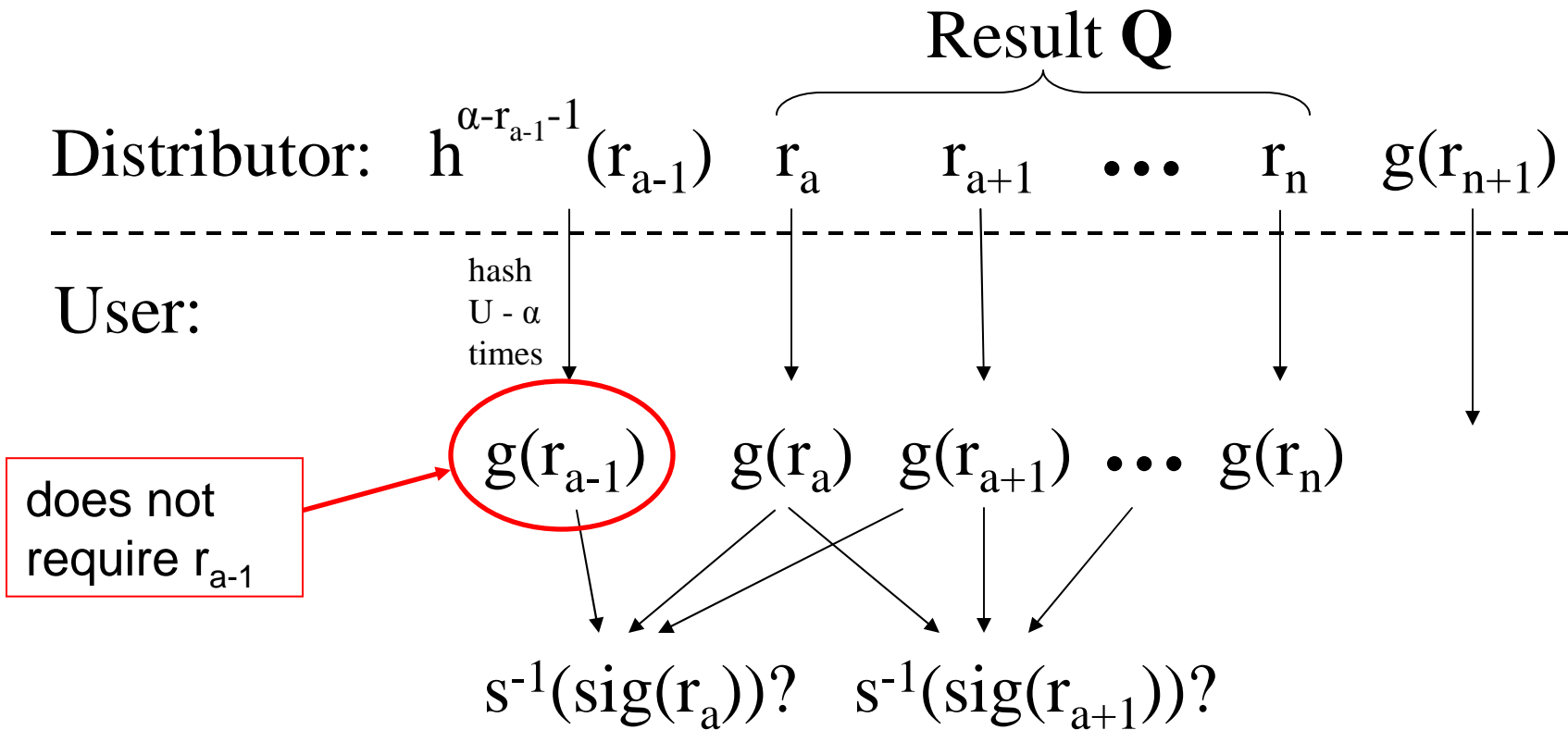
Introduces “precision” (only data
matching the query should be returned)

Idea: use signature chains – thus no need to reveal boundary elements.

$$\text{sig}(r_i) = s(\text{h}(g(r_{i-1}) \mid g(r_i) \mid g(r_{i+1})))$$



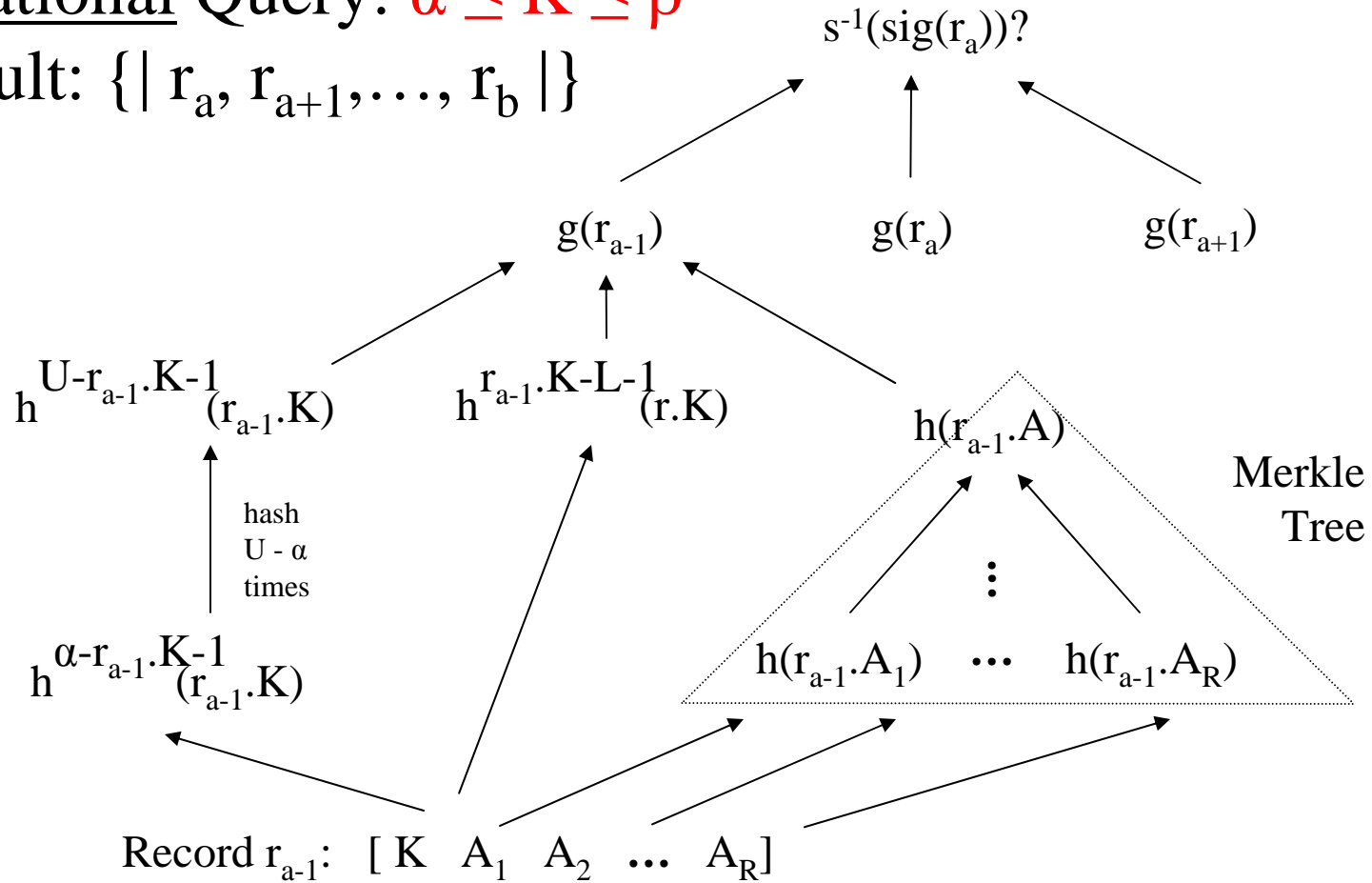
But what is g : $g(r) = h^{U-r-1}(r)$



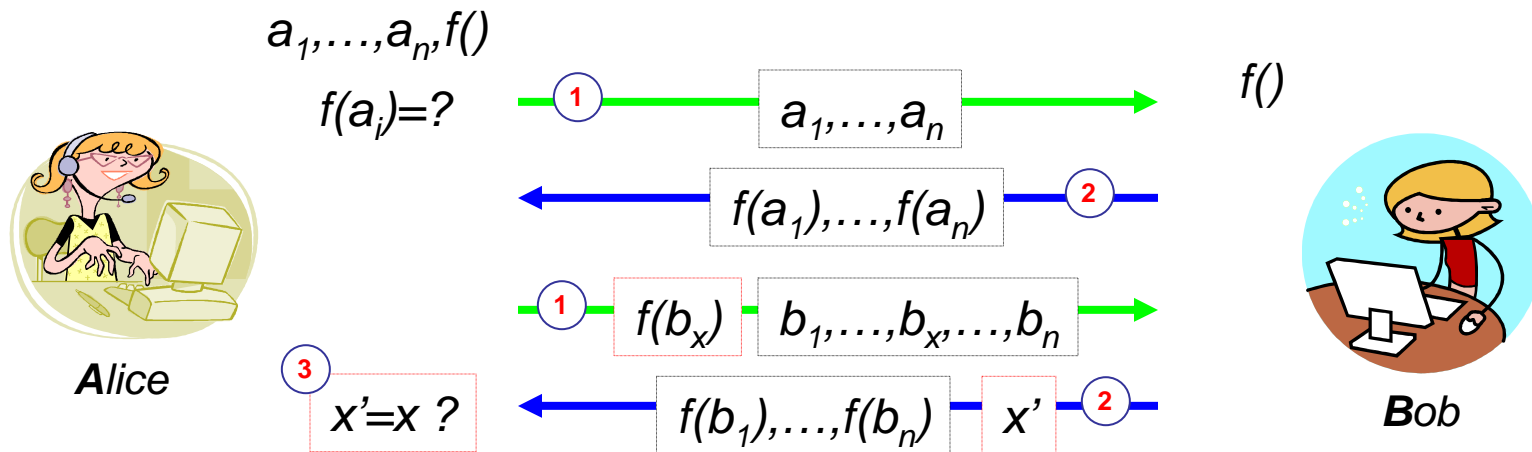
Query: $\alpha \leq r$

Relational Query: $\alpha \leq K \leq \beta$

Result: $\{ | r_a, r_{a+1}, \dots, r_b | \}$

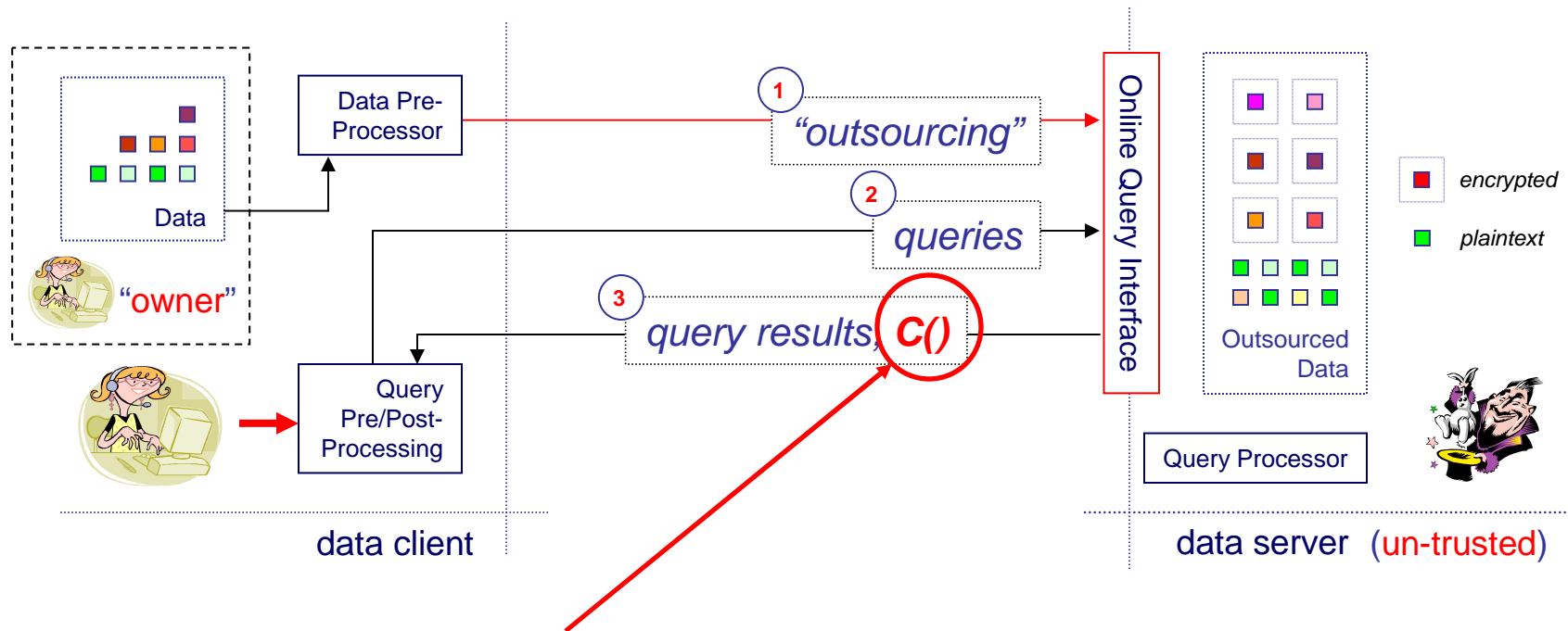


Asks: What about arbitrary queries ?



P. Golle and I. Mironov, "Uncheatable Distributed Computations", RSA 2001 (Cryptographer's track)

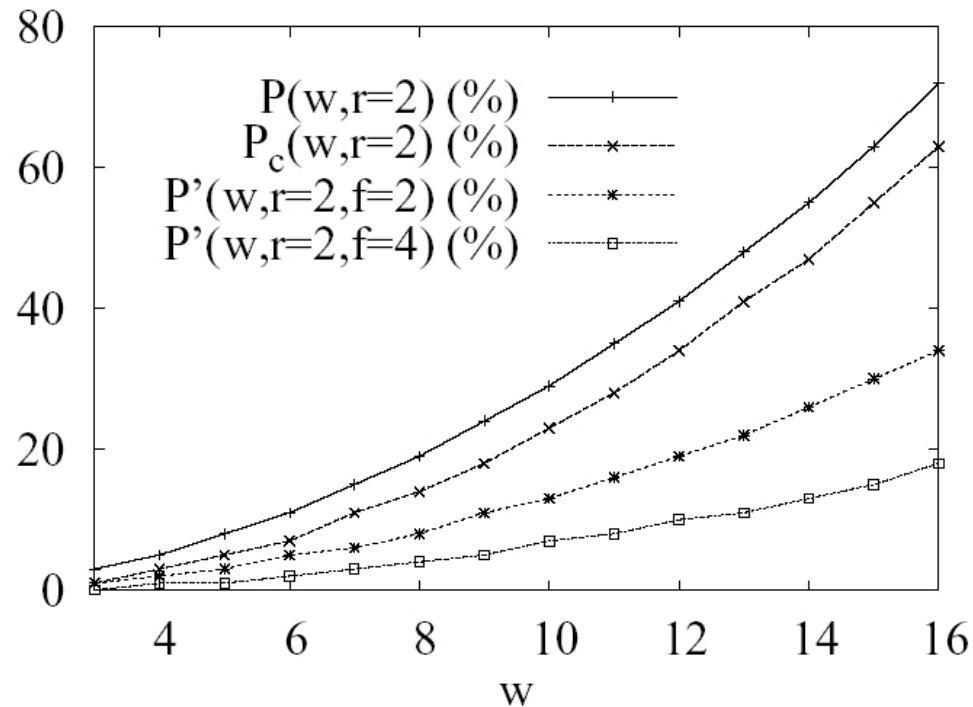
Sion: Execution Proofs



$$C(Q, x, \epsilon) = \{H(\epsilon || \rho(Q_x)), \epsilon\}$$

A challenge token (computed by client) is sent together with the batch of queries. Upon return, batch execution is proved if $x=x'$.

Sion: Cheating Probability

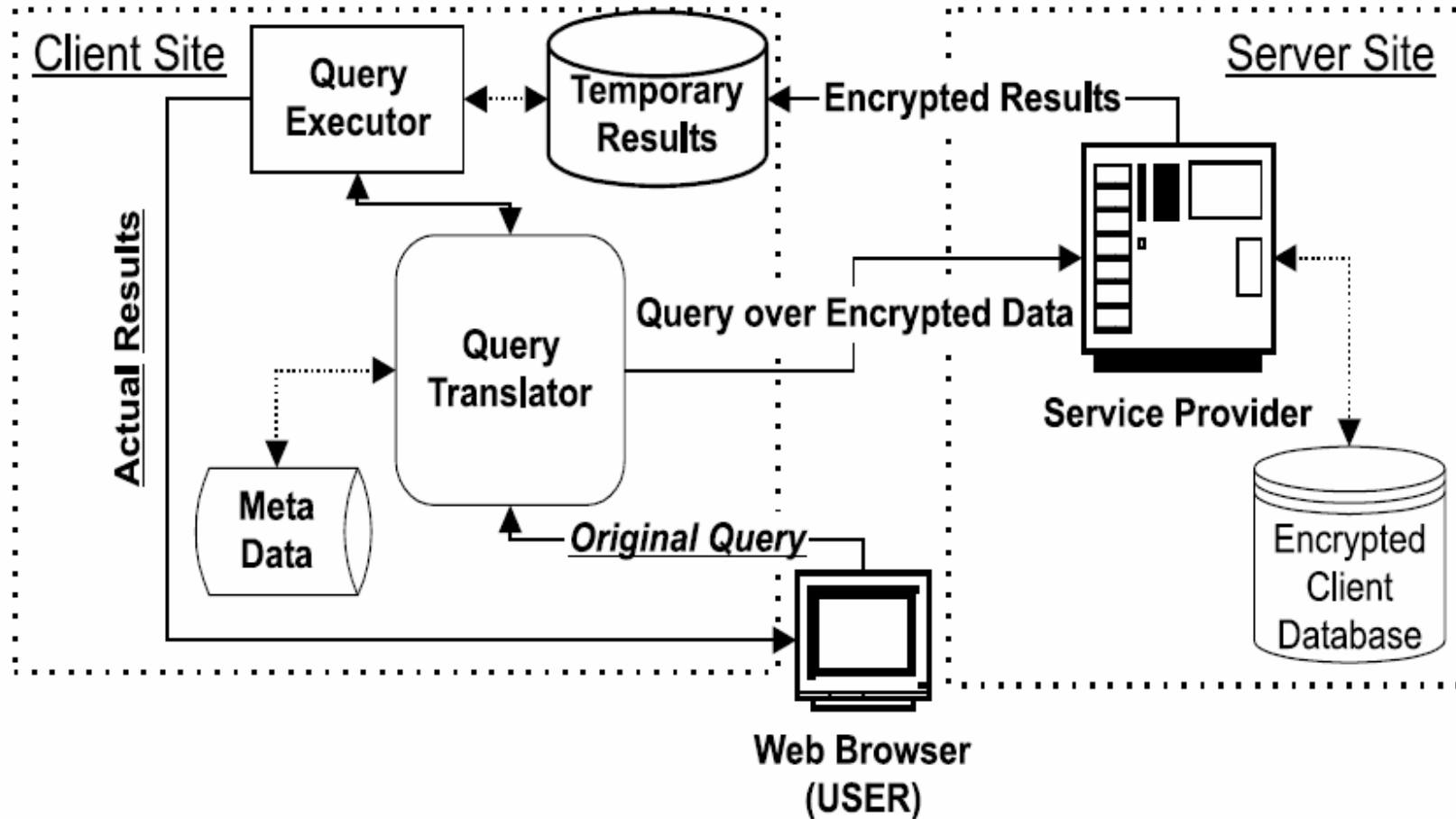


Only handles lazy server !

The behavior of $P'(w, r, f)$ (fake tokens) plotted against $P_c(w, r)$ (client-side result checking mechanism) showing that the query execution proof mechanism (with fake tokens) significantly decreases the ability to “get away” with less work.

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Hacigumus (SIGMOD 2002)



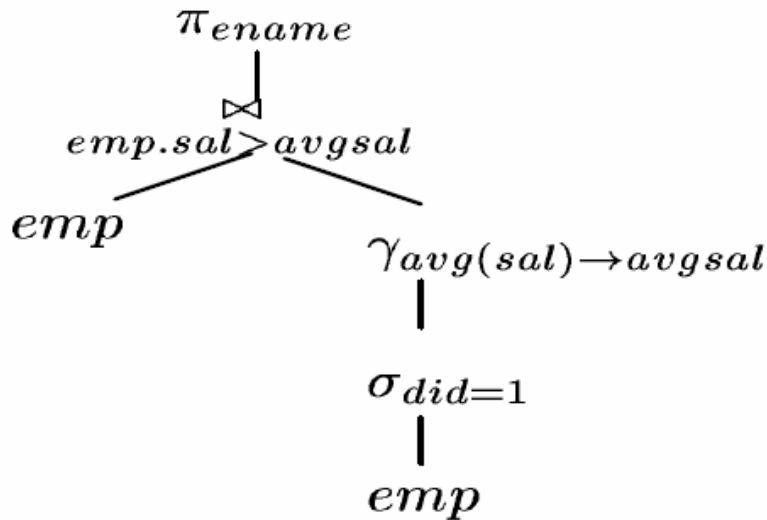
Main Steps:

1. Partition sensitive domains
 - Order preserving: supports comparison
 - Random: query rewriting becomes hard
2. Rewrite queries to target partitions
3. Execute queries and return results
4. Prune/post-process results on client

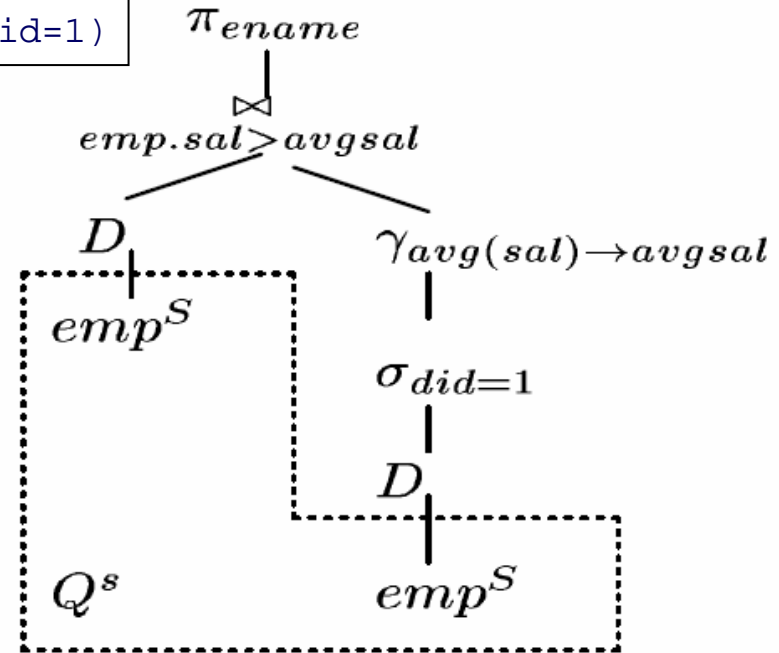
Hacigumus (SIGMOD 2002)

```

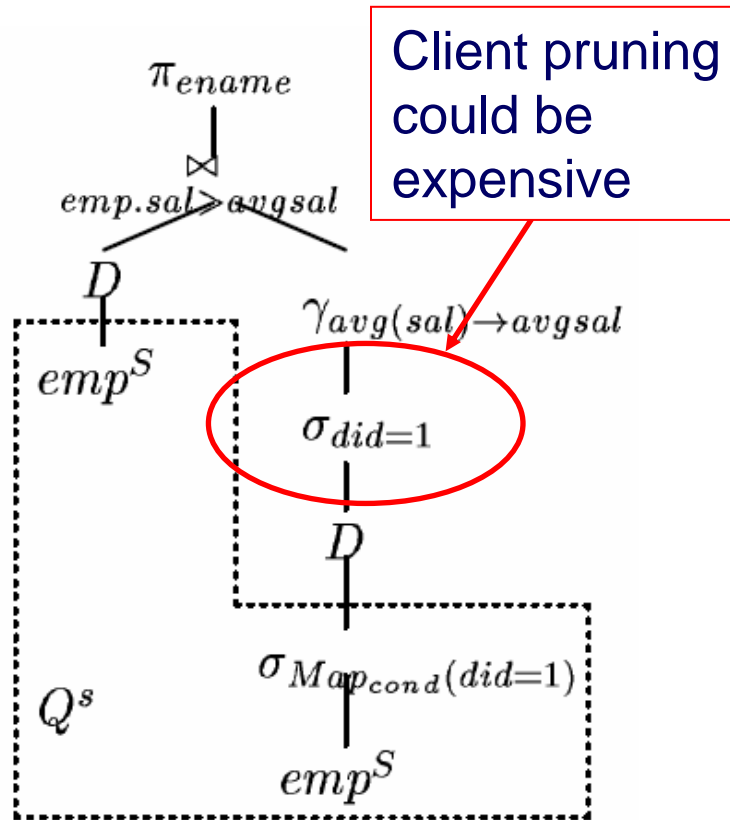
SELECT emp.name FROM emp
WHERE emp.salary >
  (SELECT AVG(salary) FROM emp WHERE did=1)
    
```



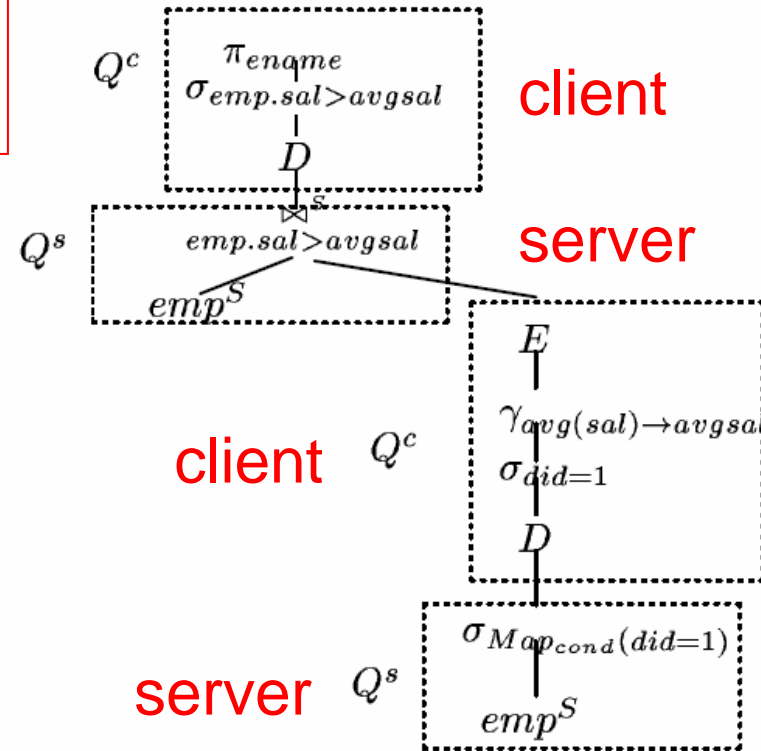
(a) Original query tree.



(b) Replacing encrypted relations.



(c) Doing selection at server.



(d) Multiple interactions between Client and Server.

Confidentiality-Overhead Trade-off

Larger segments ==
increased privacy ==
increased overheads

Goal: For a uniform distribution of queries
- minimize any leaks to any adversaries
(even) knowing segmentation parameters.

Idea 1: Maximize variance of distribution
of values in segment

Idea 2: Increase segment entropy

Issue: What about performance ?

Solution: “Controlled Diffusion”

Idea:

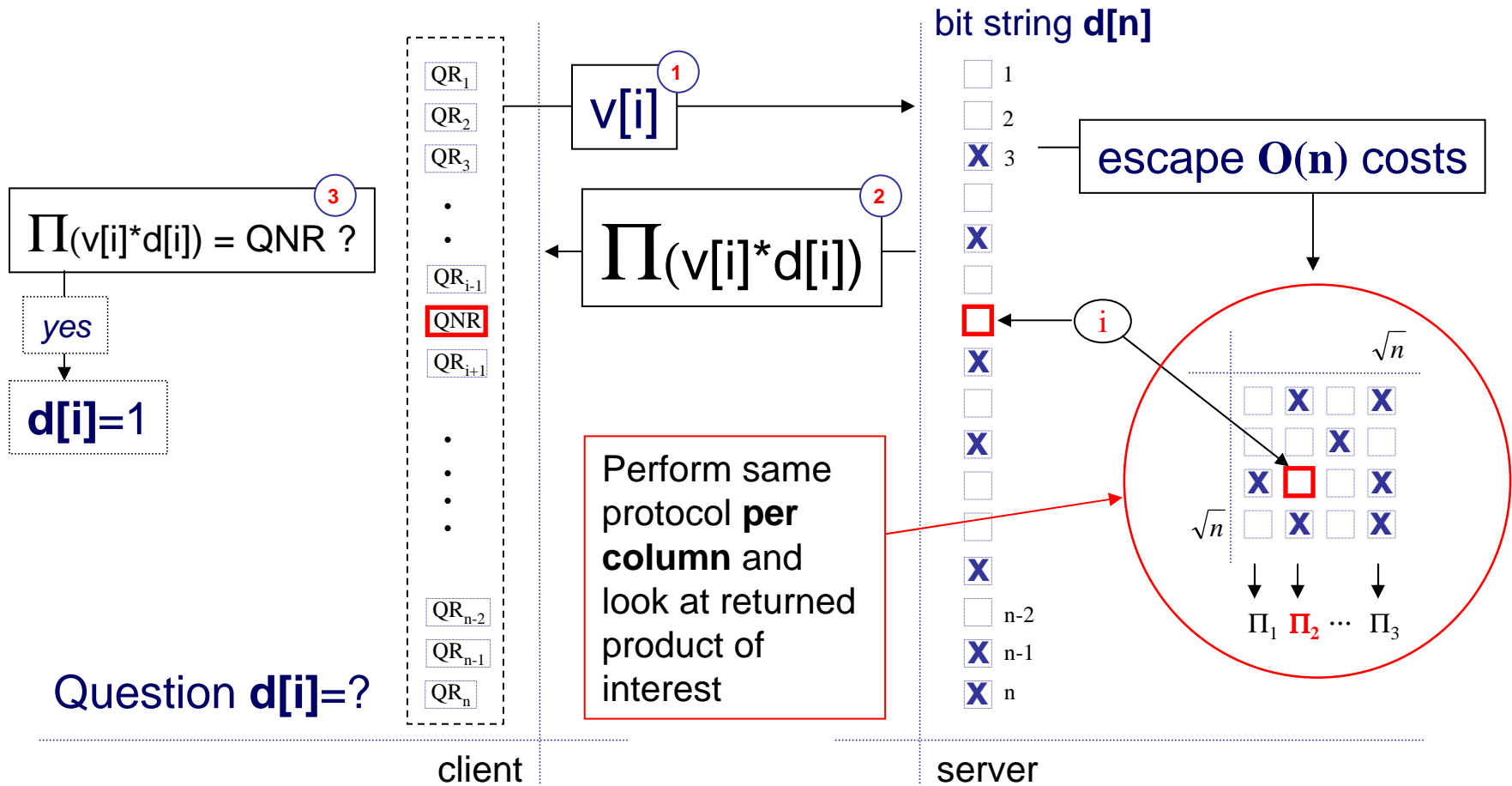
1. design for efficiency, then ...
2. ... diffuse (re-distribute) elements inside the segments to increase per-segment entropy and variance

Asks: Similarly, how to structure query trees to optimally balance the security-efficiency trade-off in [Hacigumus 2002].

Idea: client generates optimal partitioned query execution plans given statistics and metadata input from the server.

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QR PIR



The n bits of the database are organized logically at the server as a bi-dimensional matrix M of size $\sqrt{n} \times \sqrt{n}$. To retrieve bit $M(x, y)$ with computational privacy, the client:

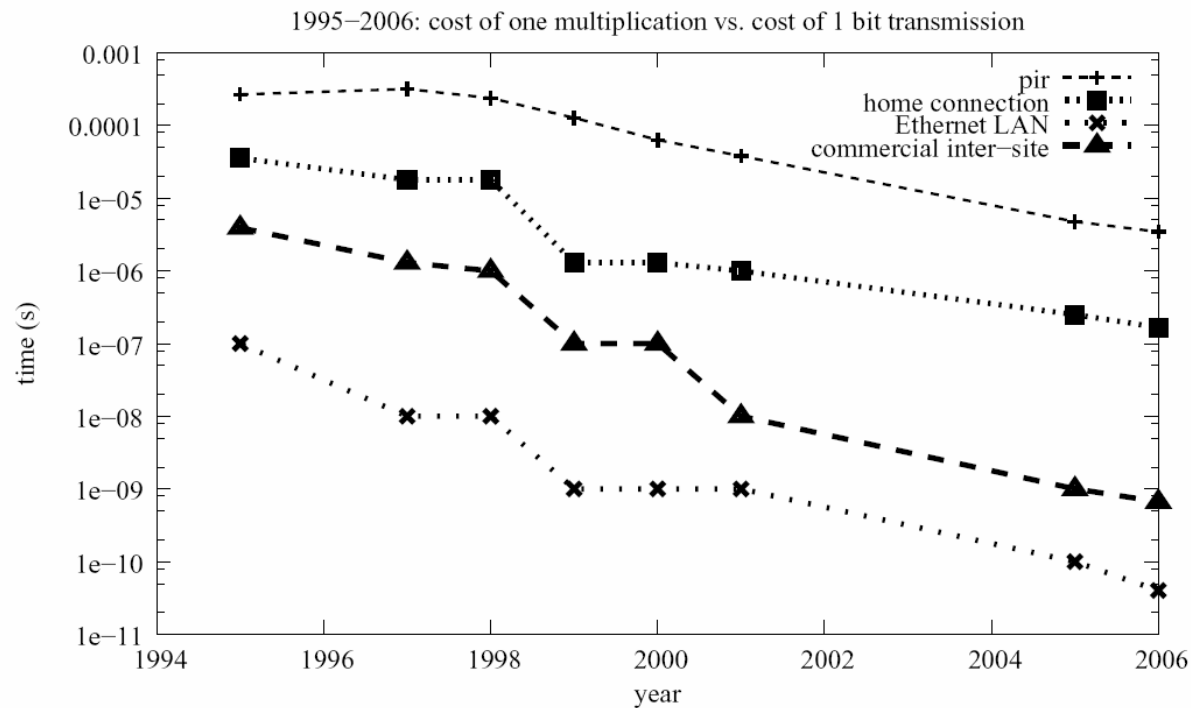
- randomly chooses two prime numbers p and q of similar bit length, computes their product, $N = pq$ and sends it to the server.
- generates \sqrt{n} numbers $s_1, s_2, \dots, s_{\sqrt{n}}$, such that s_x is a quadratic non-residue (QNR) and the rest are quadratic residues (QR) in \mathbb{Z}_N^* .
- sends $s_1, s_2, \dots, s_{\sqrt{n}}$ to the server.

For each “column” $j \in (1, \sqrt{n})$ in the $\sqrt{n} \times \sqrt{n}$ matrix, the server:

- computes the product $r_j = \prod_{0 < i < \sqrt{n}} q_{ij}$ where $q_{ij} = s_i^2$ if $M(i, j) = 1$ and $q_{ij} = s_i$ otherwise ².
- sends $r_1, \dots, r_{\sqrt{n}}$ to the client

The client then simply checks if r_y is a QR in \mathbb{Z}_N^* which implies $M(x, y) = 1$, else $M(x, y) = 0$.

PIR is (still) impractical

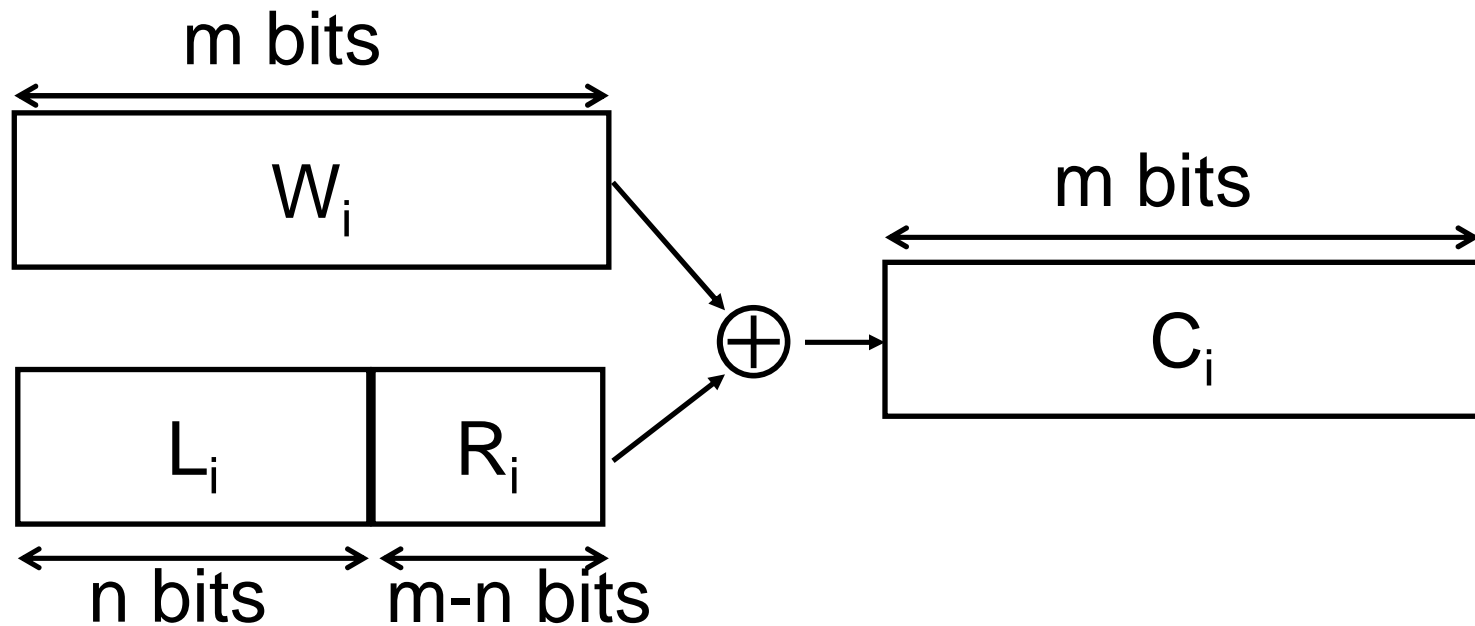


Comparison between the time required to perform PIR and the time taken to transfer the database, between 1995 and 2005. (logarithmic)

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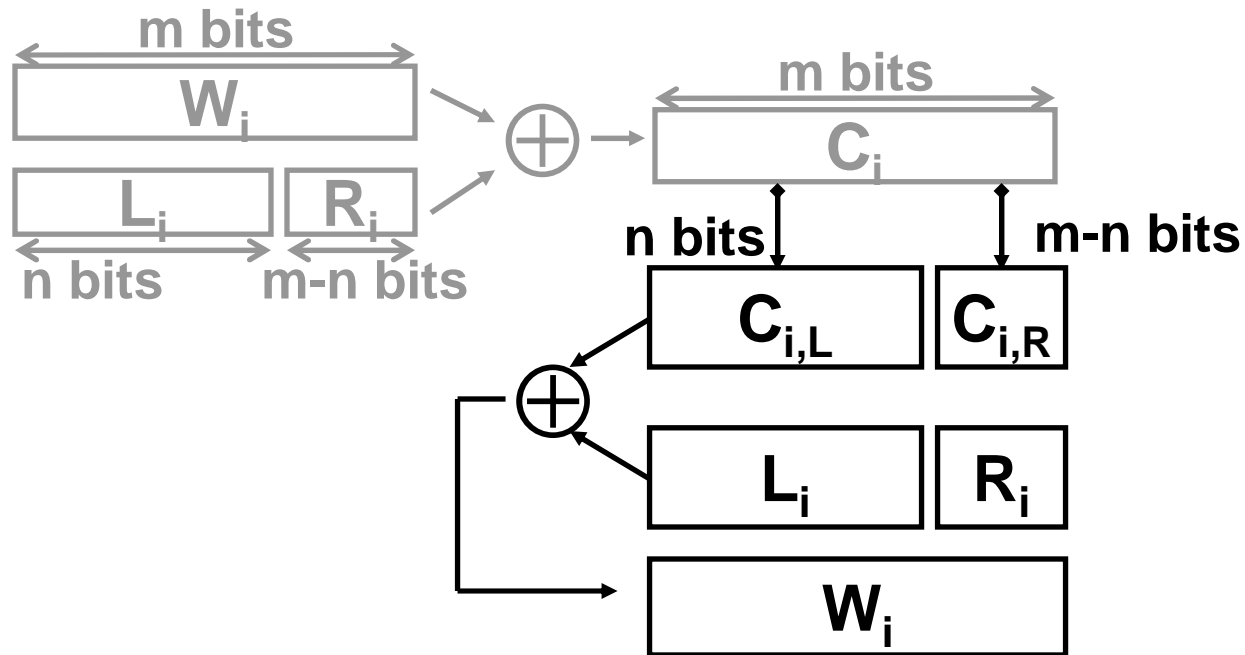
- Sequential Scan
- Index-based

Encryption:



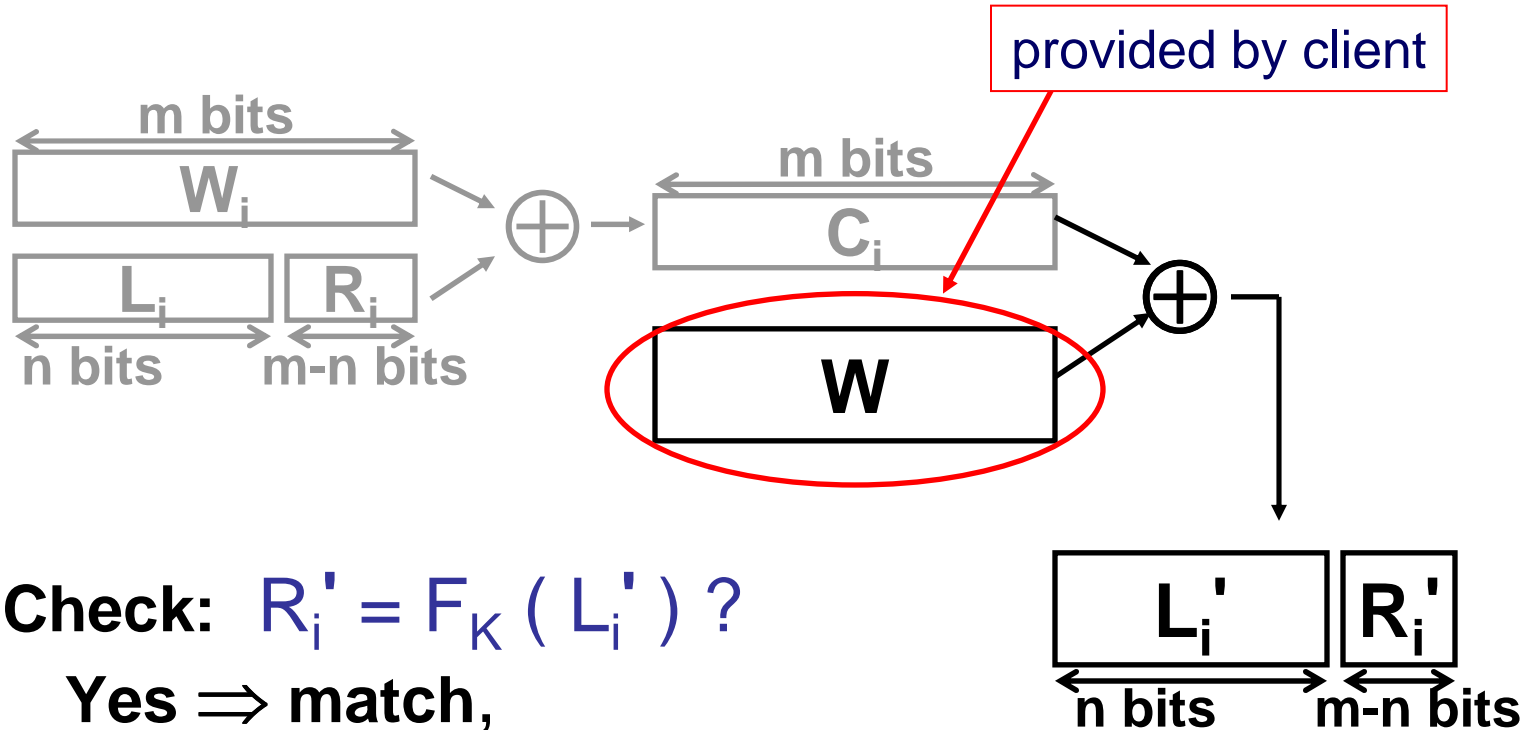
$$L_i \leftarrow G_i(\text{seed}), \quad R_i \leftarrow F_K(L_i)$$

Decryption:



$$L_i \leftarrow G_i(\text{seed}), \quad R_i \leftarrow F_K(L_i)$$

Search:

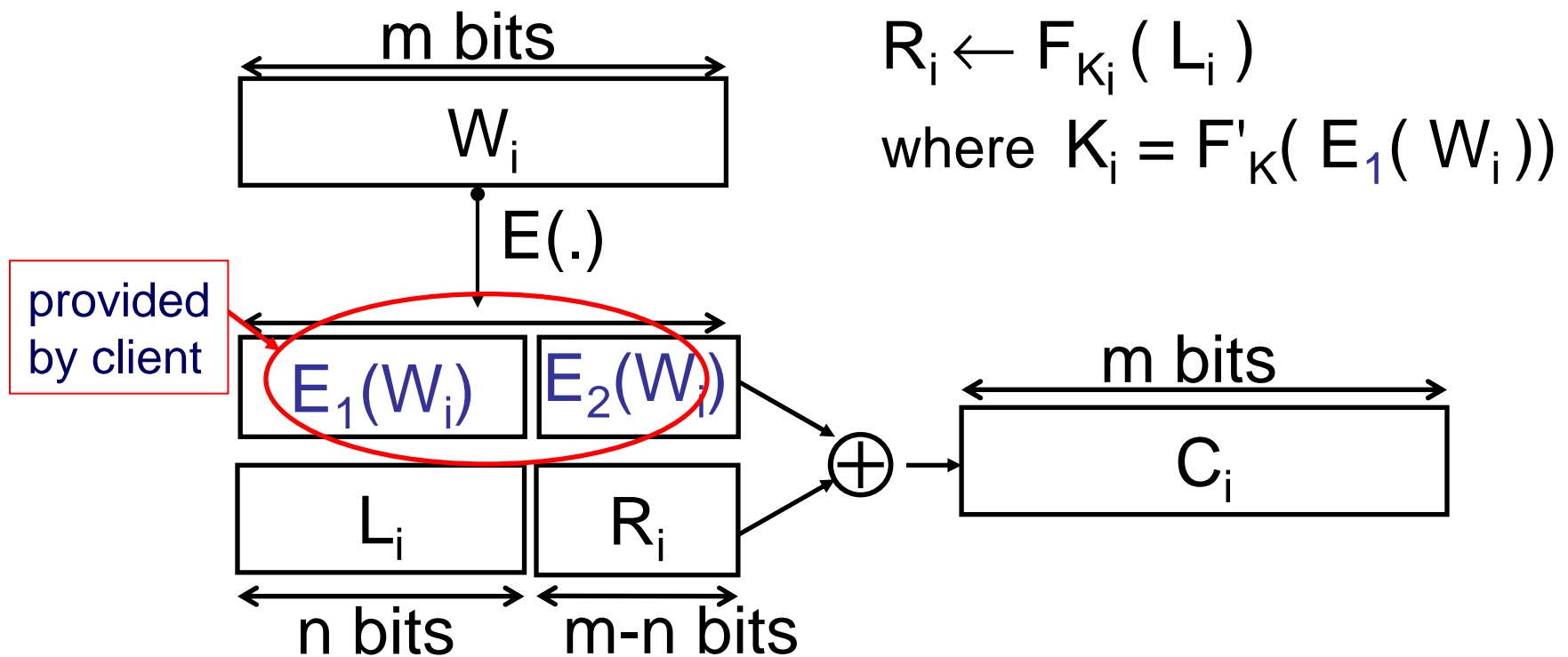


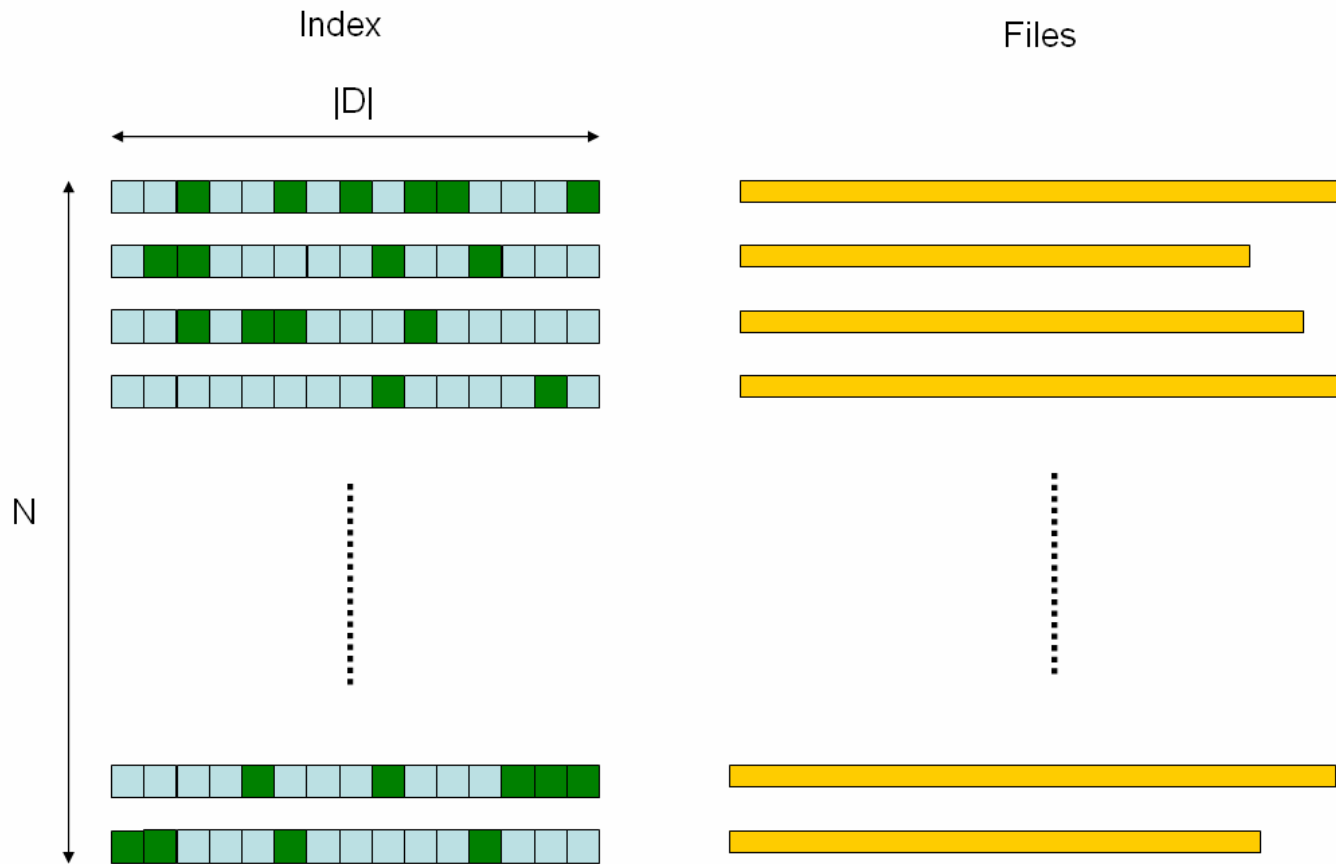
Check: $R'_i = F_K(L'_i)$?

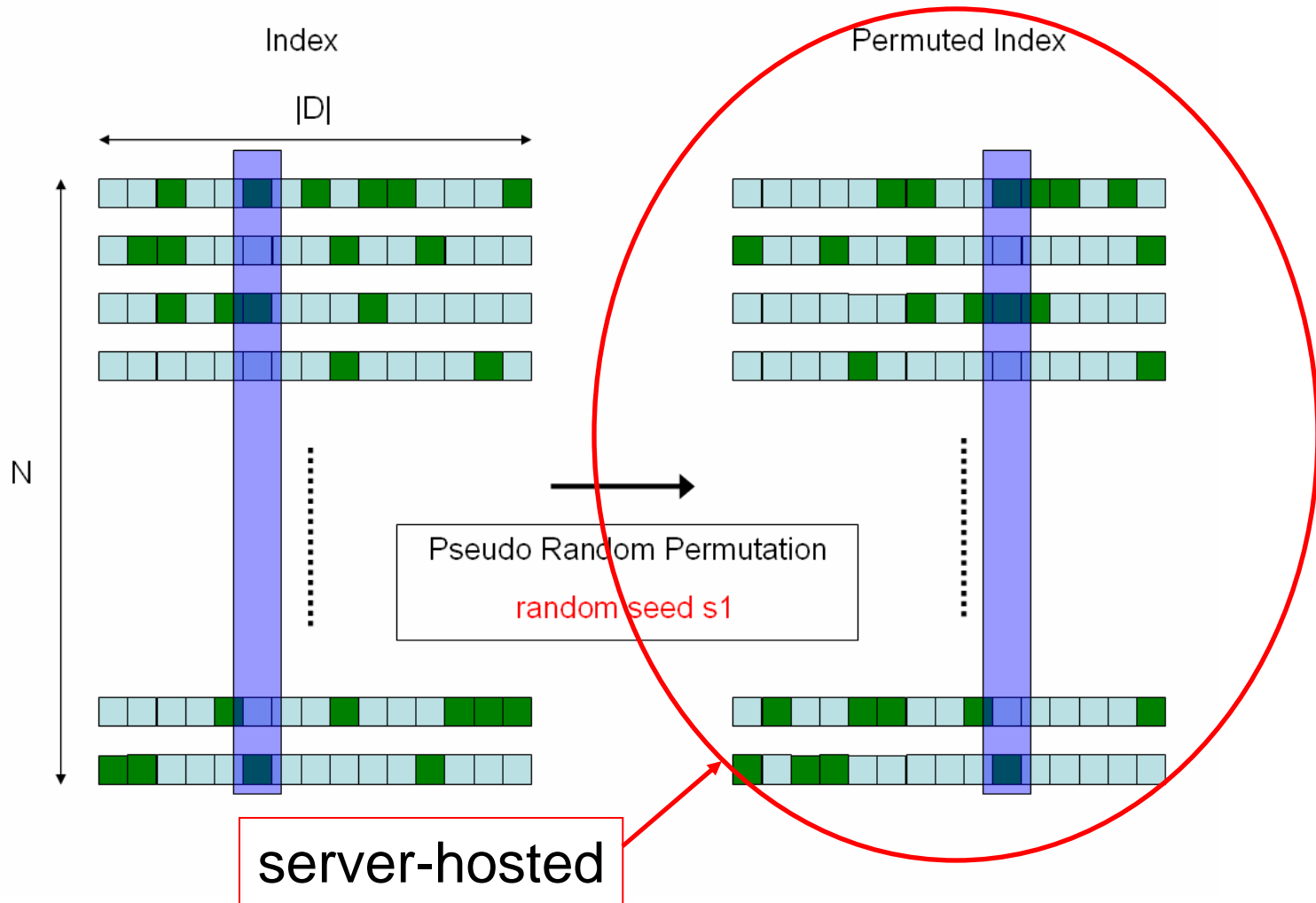
Yes \Rightarrow match,

(false positive rate = $1 / 2^{m-n}$)

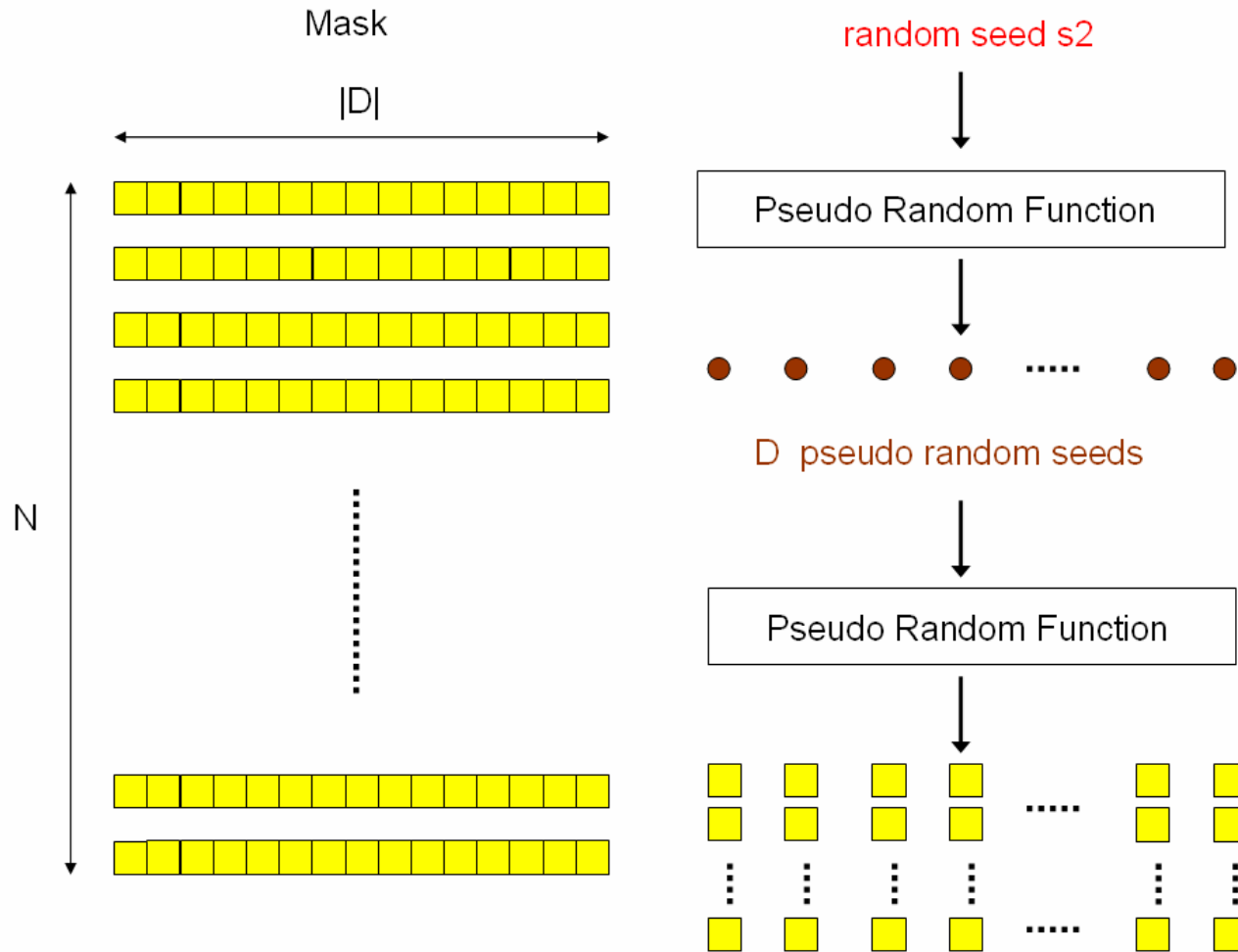
“Hidden” Search:



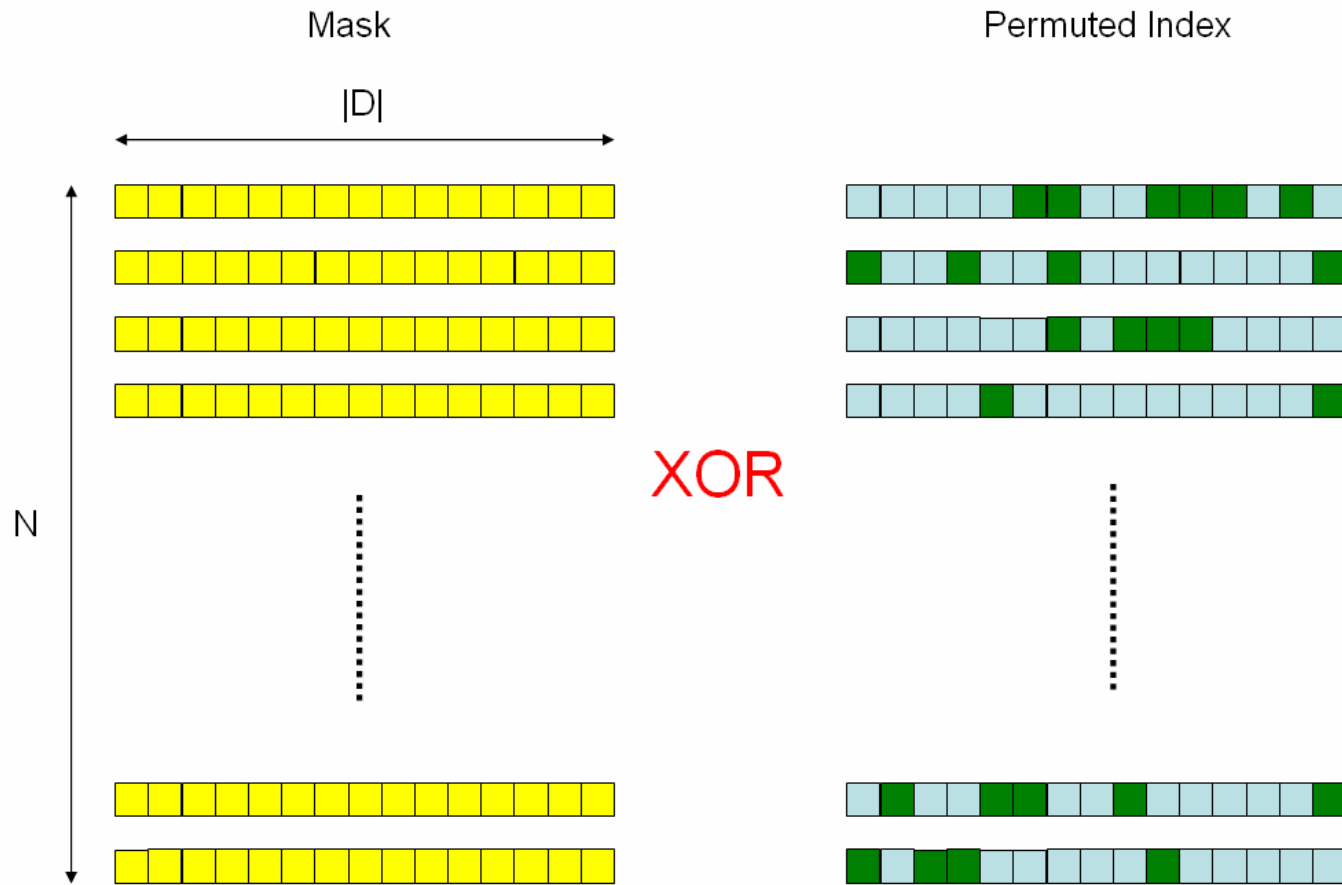


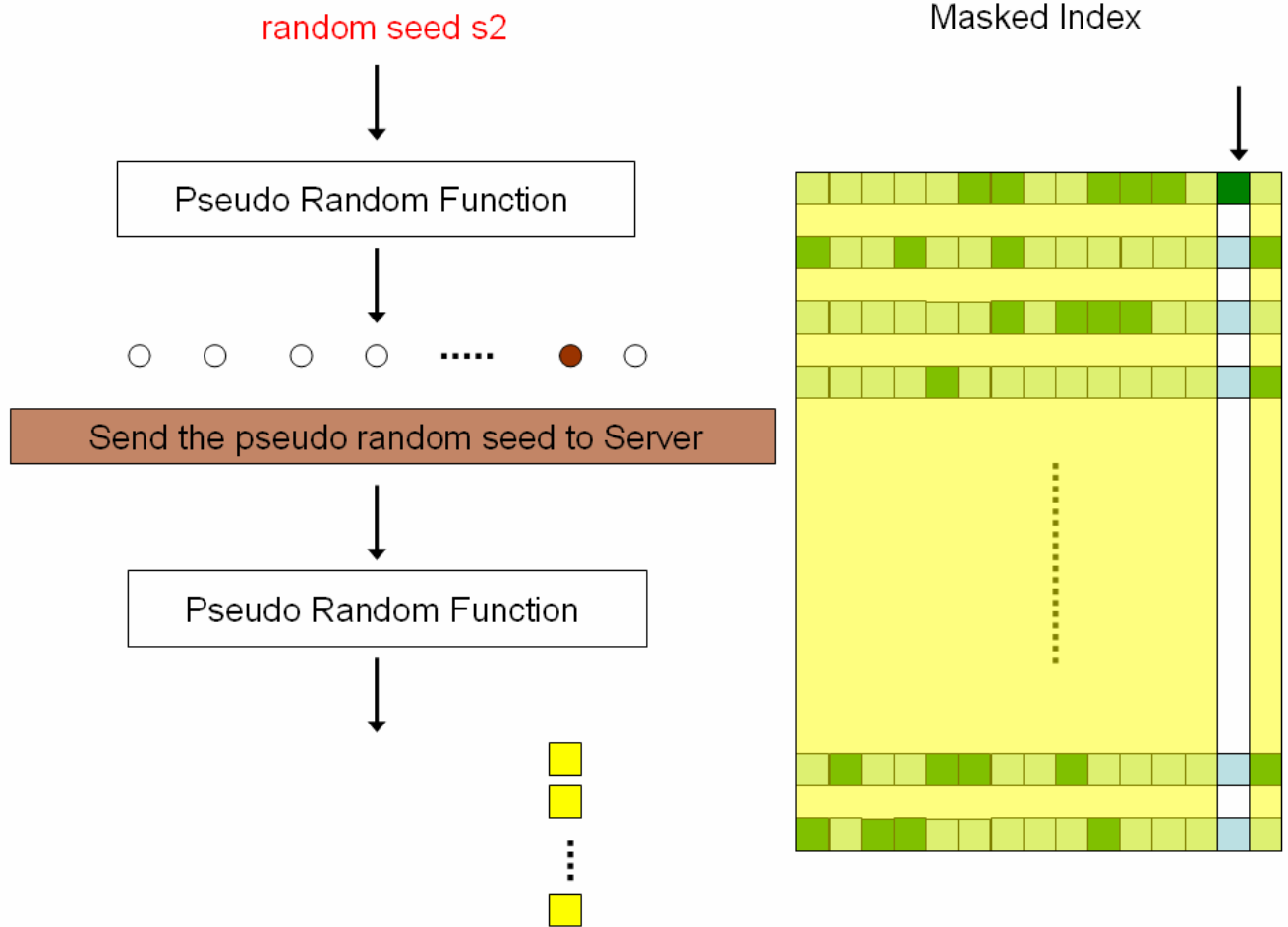


Chang (2004)



Chang (2004)





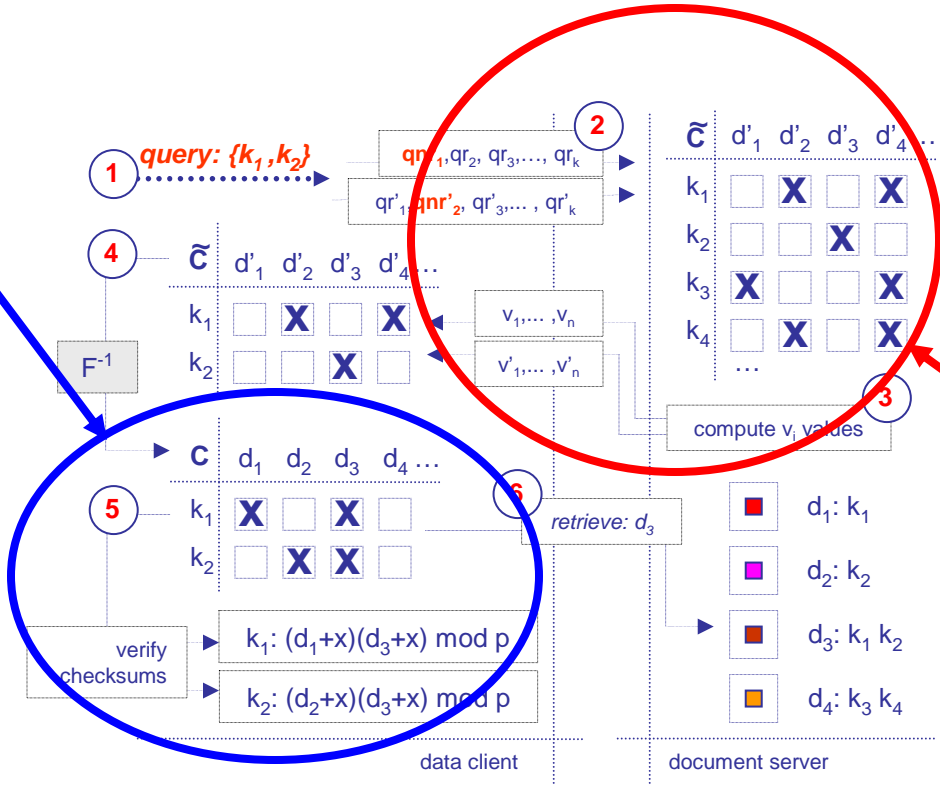
Server stores **capabilities** for conjunctive queries (linear in the total number of documents). These can be transferred offline.

The client is required to know before-hand future conjunctive queries.

Query part is sent online at the time of search. It is of constant size (number of keyword fields per documents).

Asks: What about correctness + privacy ?

Query Correctness



Computational Privacy

Idea: Deploy modified version of computational PIR targeted at a server-side index. Augment with “multiplicative checksums”.

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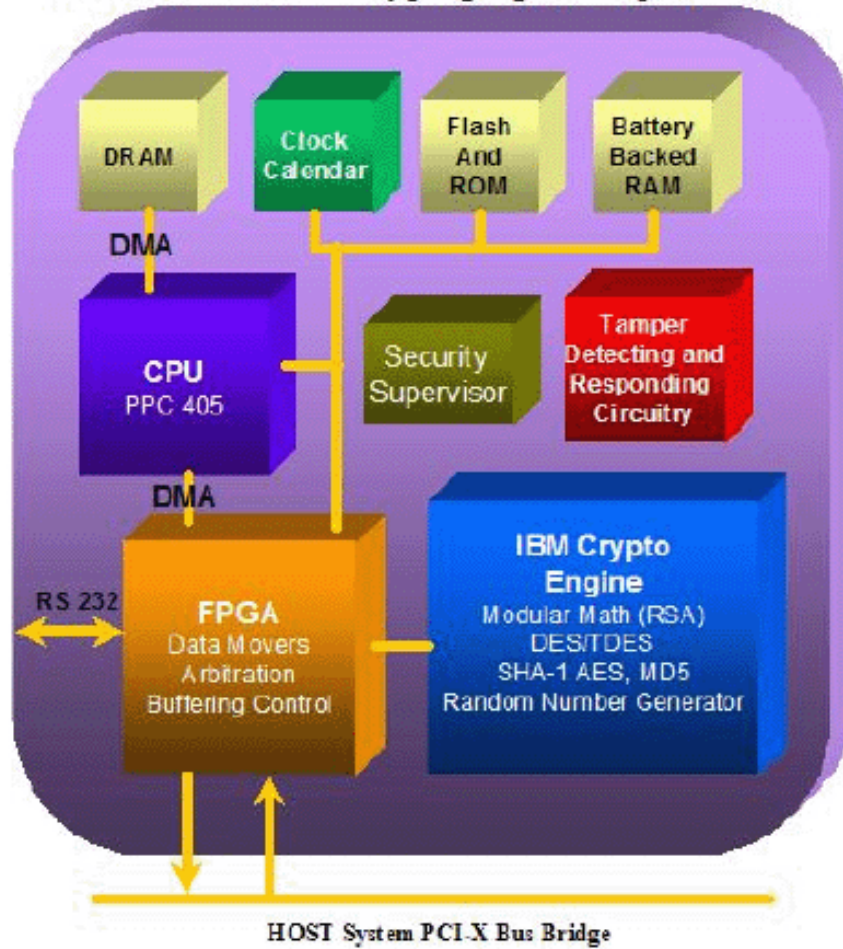
Trusted Hardware



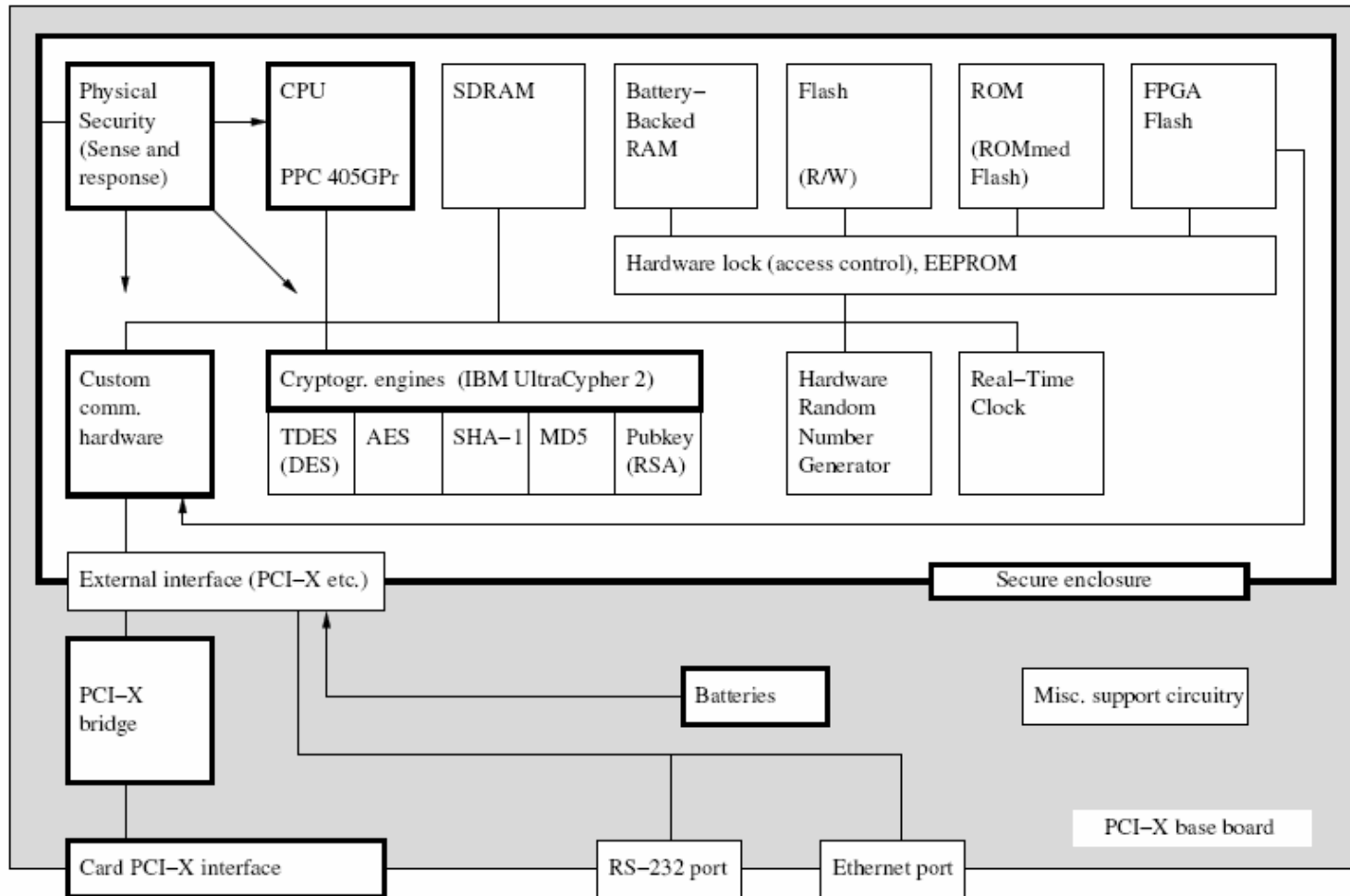
IBM 47xx

IBM 4764 Architecture

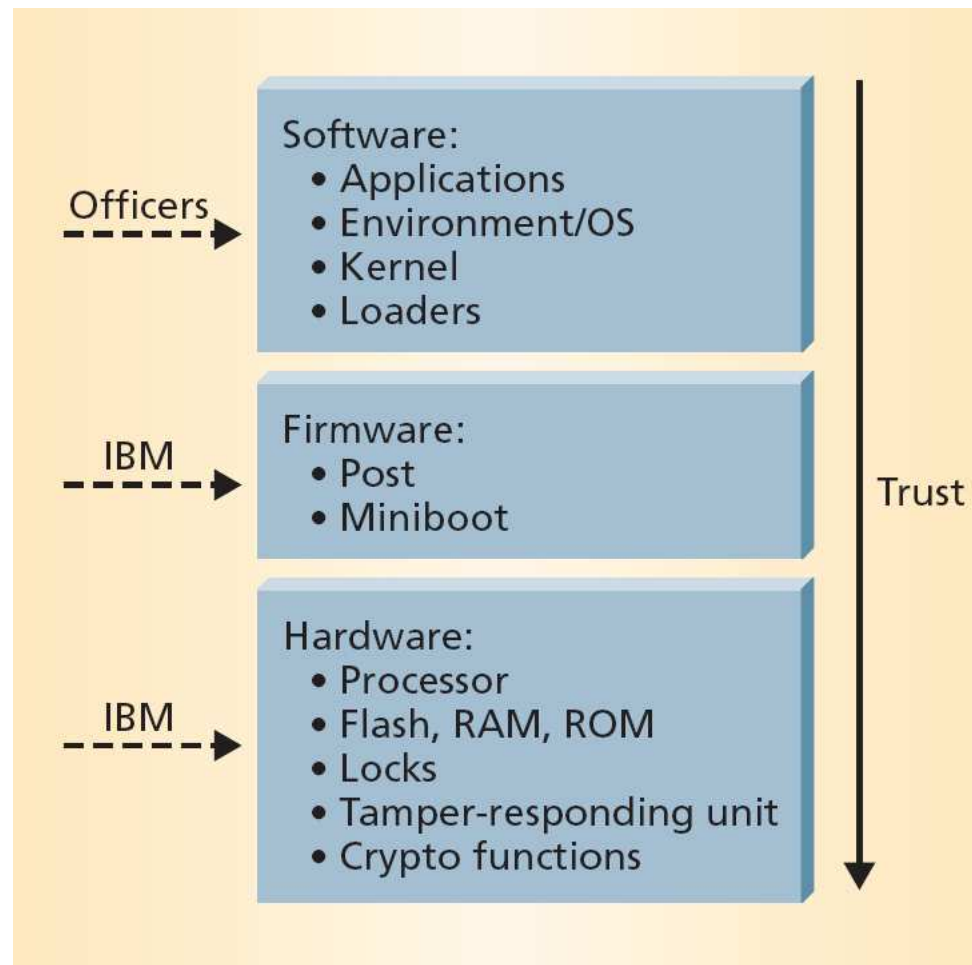
IBM 4764 PCI-X Cryptographic Coprocessor



IBM 4764 Architecture



Trust Propagation



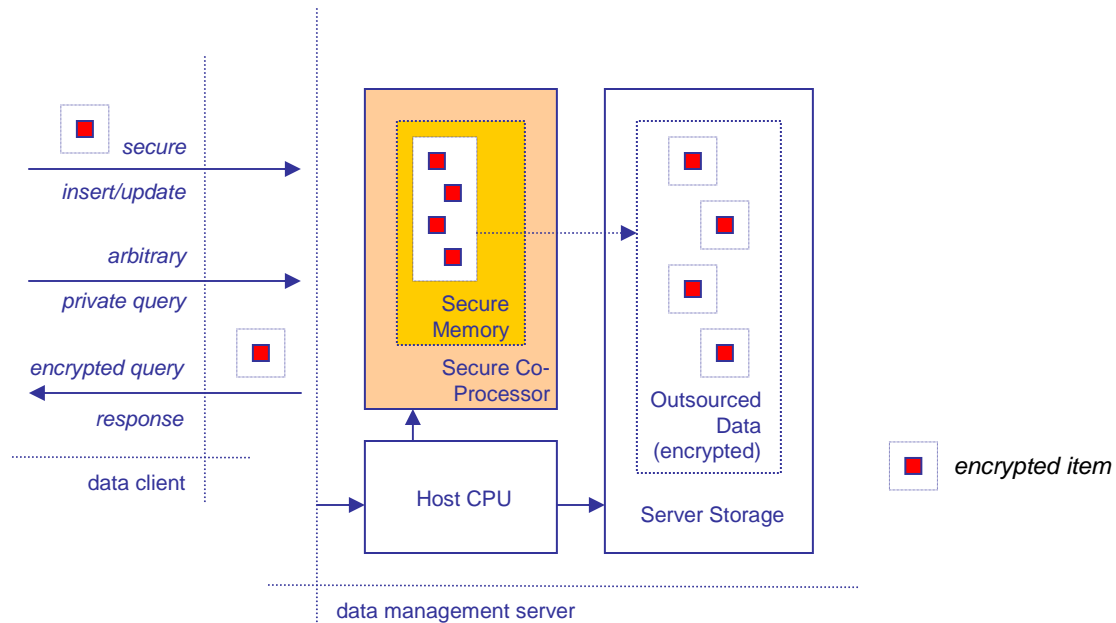
SCPU Performance



RSA1024 Sign: 848/sec
RSA1024 Verify: 1157/sec
3DES: 1-8MB/sec
DES: 1-8MB/sec
SHA1: 1-21MB/sec

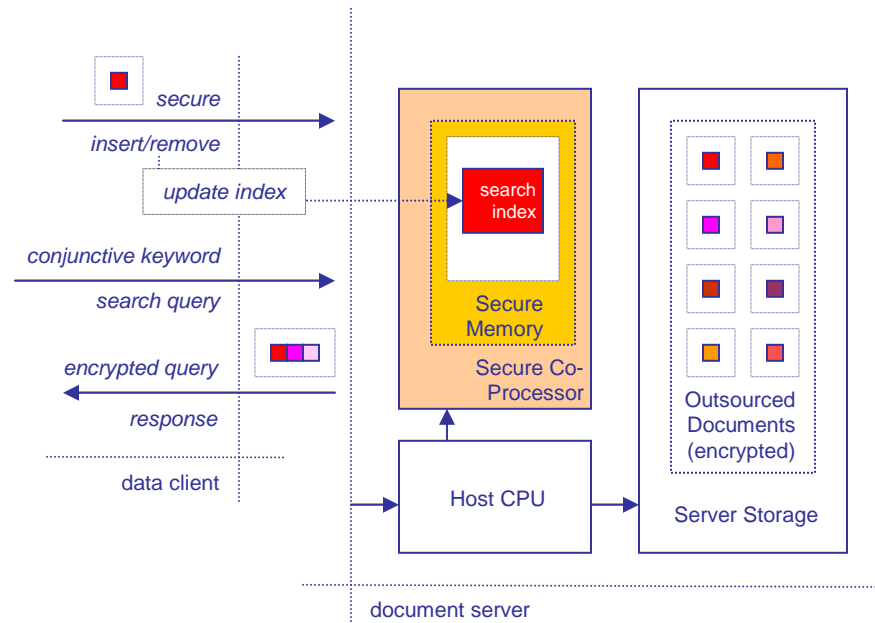
IBM 4764-001: 266MHz PowerPC. 64KB battery-backed SRAM storage. Crypto hardware engines: AES256, DES, TDES, DSS, SHA-1, MD5, RSA. FIPS 140-2 Level 4 certified.

Possible Benefits



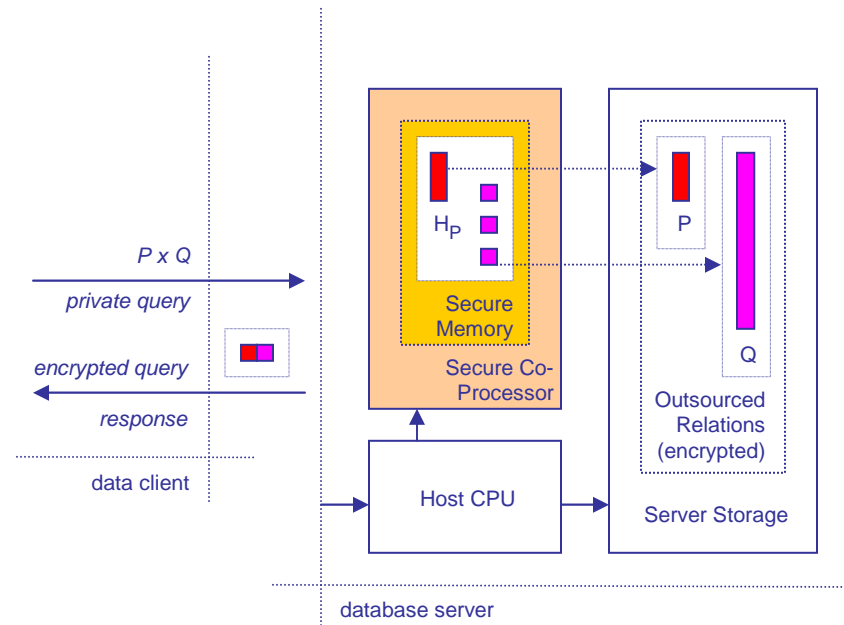
A secure co-processor on the data management side may allow for significant leaps in expressivity for queries where privacy and completeness assurance are important.

Searching



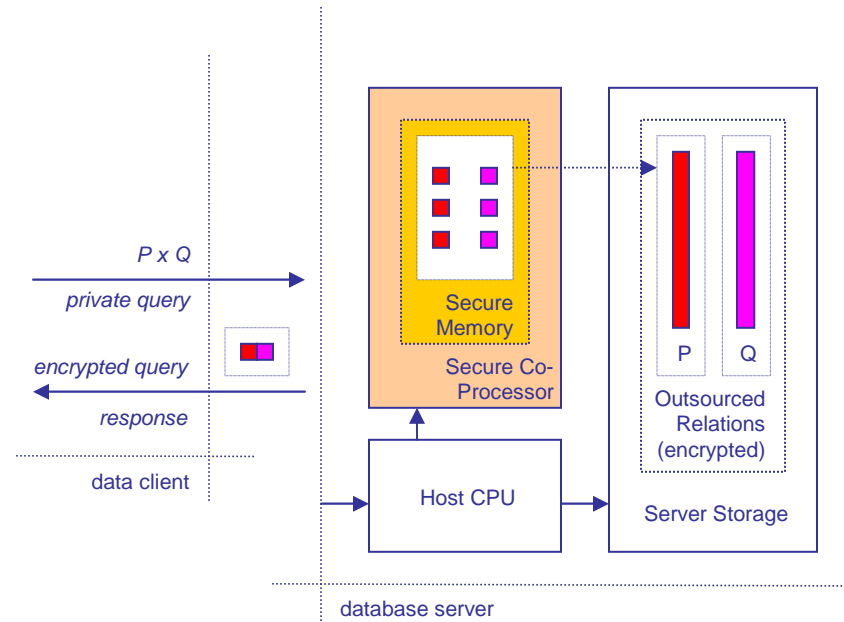
For conjunctive keyword searches on document (email, files) servers, oblivious search index structures could be queried in secure memory achieving a novel zero-leak query model.

Hash-JOIN



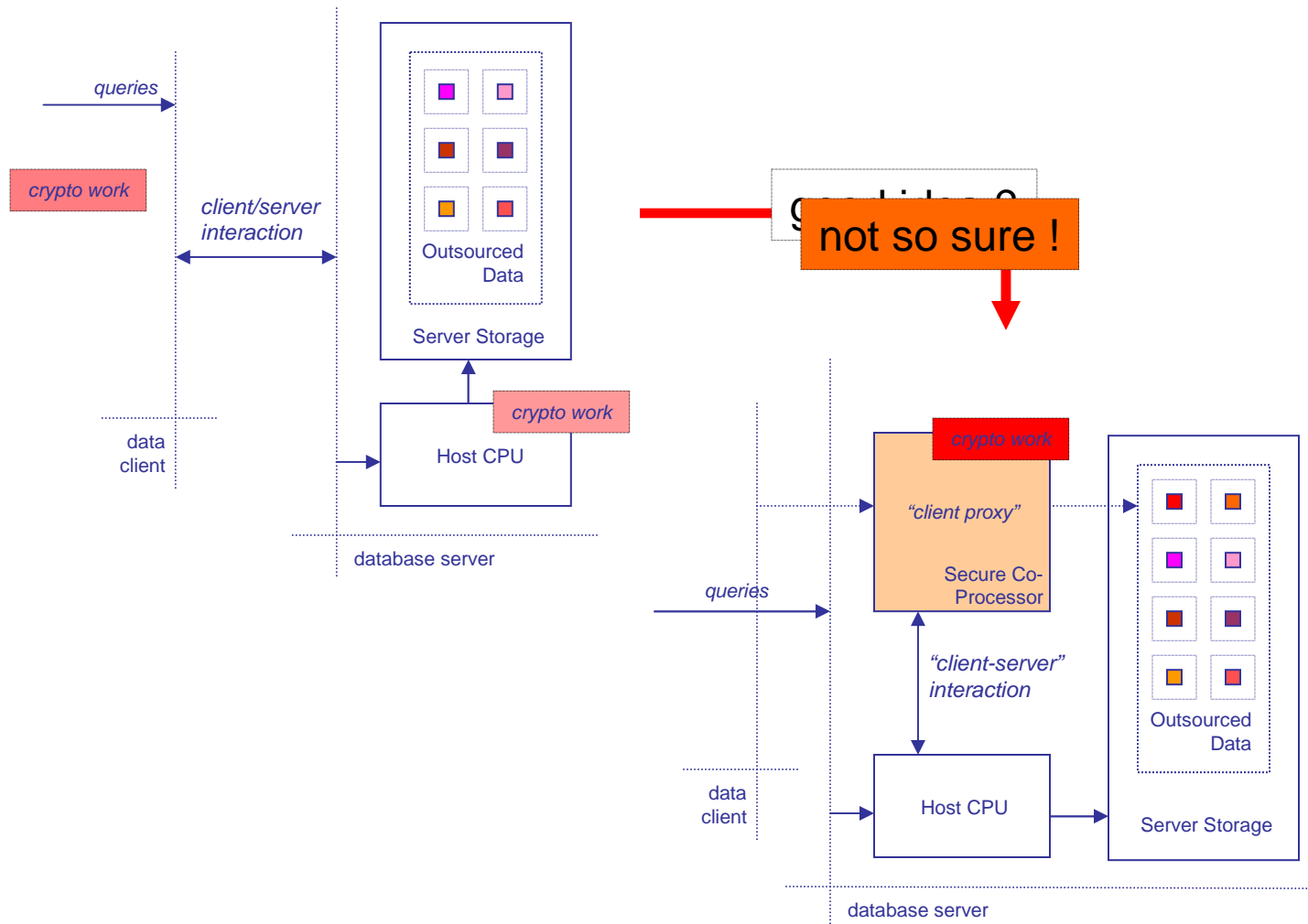
Hash-JOIN could be naturally accommodated.

Merge-JOIN

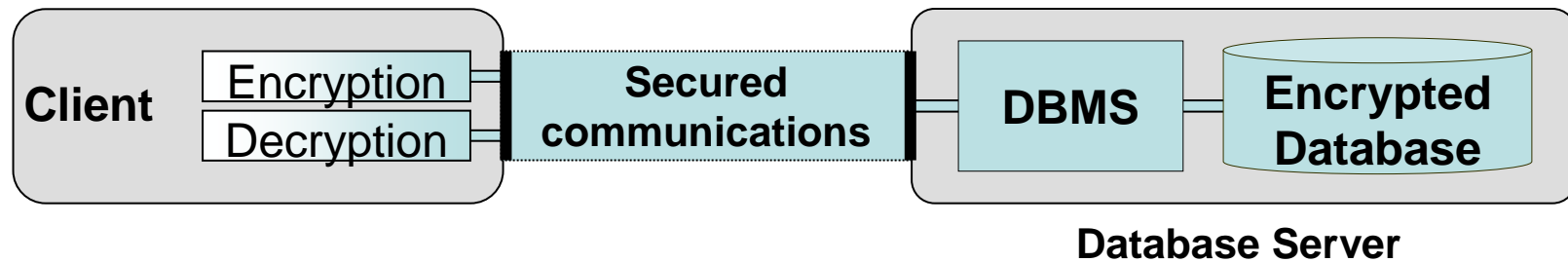
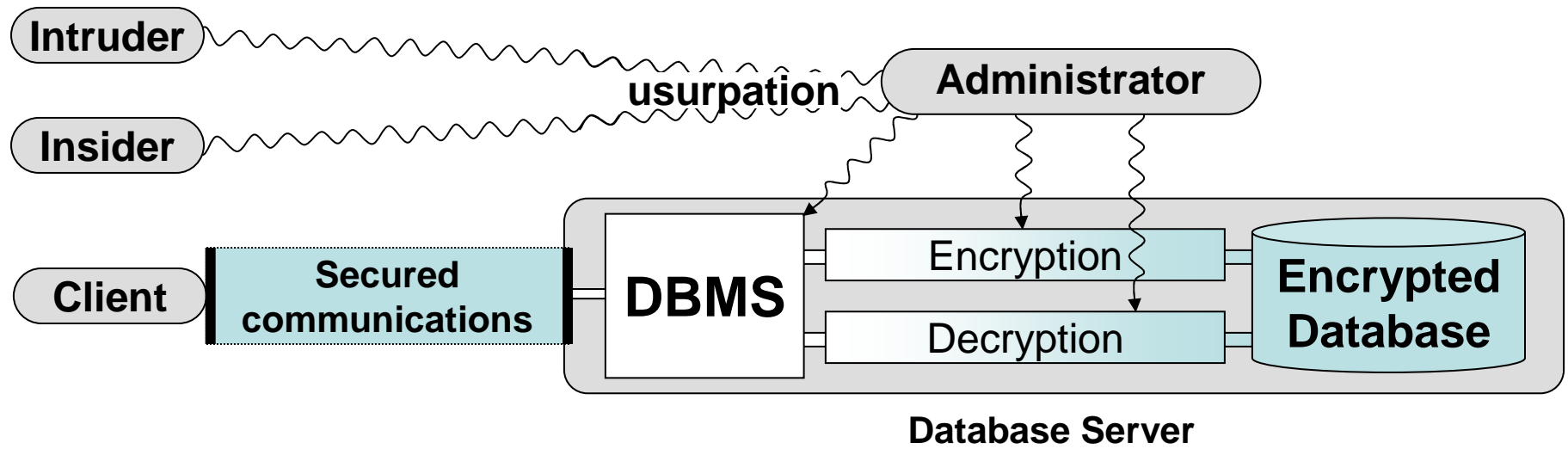


For Merge-JOIN, order-preserving encryption primitives could be deployed to minimize the amount of data parsing required in the sorting phase.

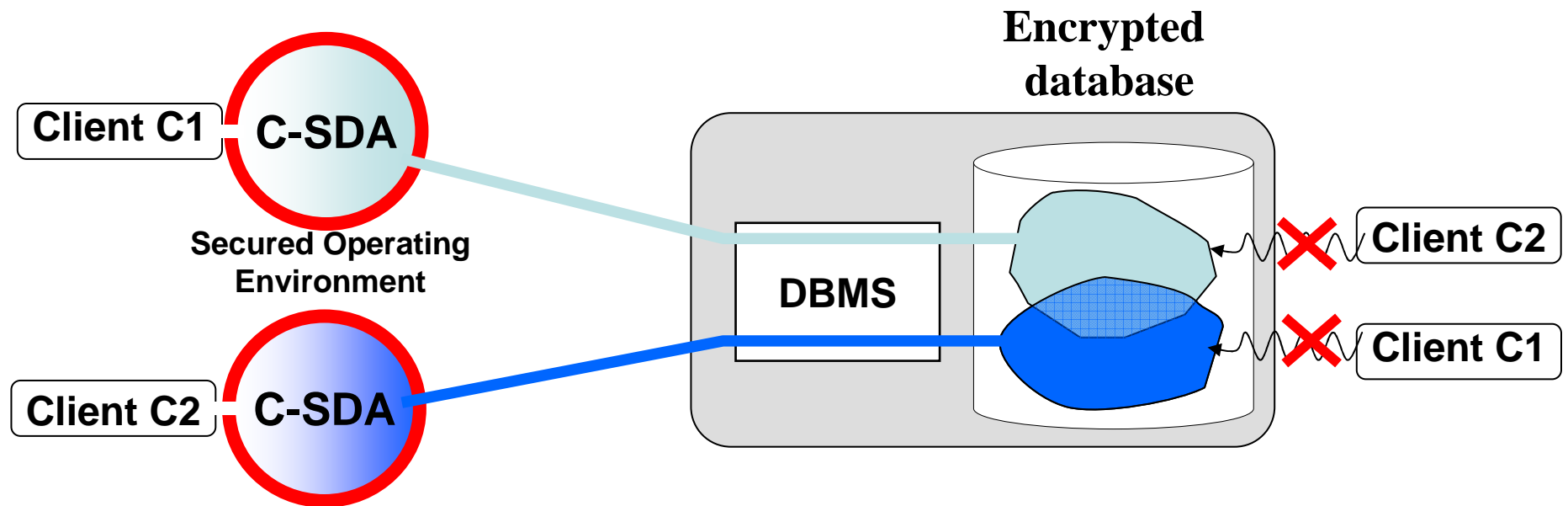
Sample DON'T



- Process entire queries on SCPU (!)
- Dedicate (one) SCPU per query or equivalent
 - e.g., limit TPS by SCPU TPS
- Synchronize CPU with SCPU
 - e.g., block main CPU until SCPU completes
- Transfer $\geq O(n)$ on SCPU-CPU bus (!)
- Anything else un-smart 😊

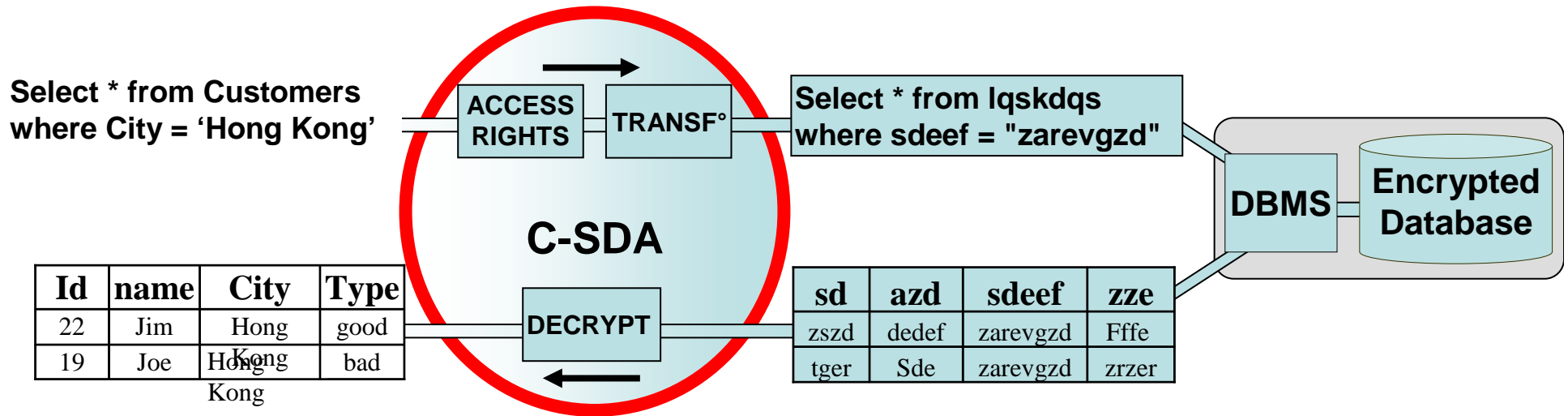


Chip-Secured Data Access:

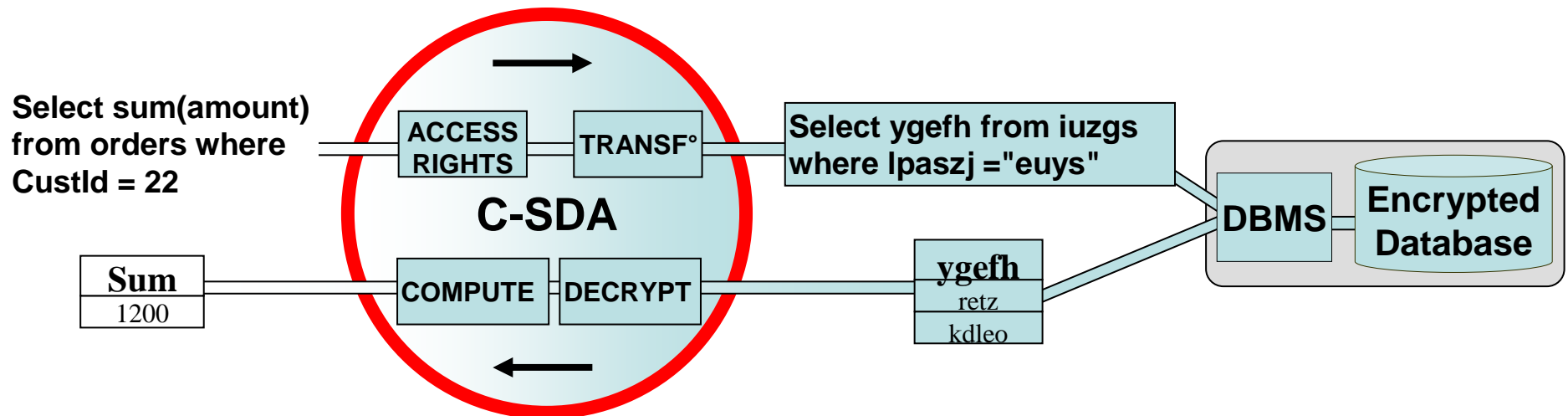


Smartcard: 32 bit RISC processor ($\approx 40\text{Mips}$), limited communication bandwidth (10 to 100 Kbps), tiny RAM, writes in EEPROM very costly.

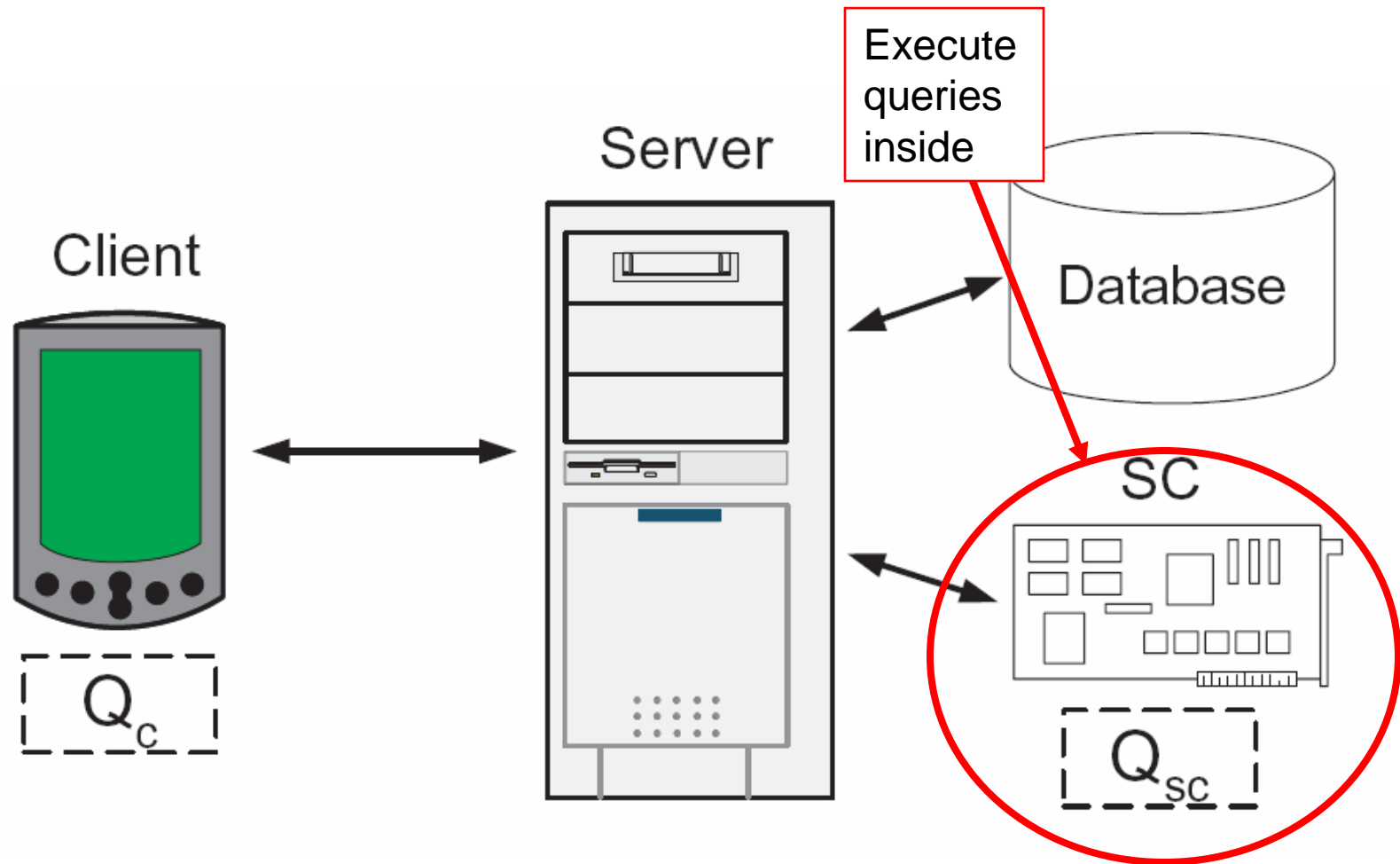
Equi-predicate-only Queries:



General queries:



Tsudik (2005)



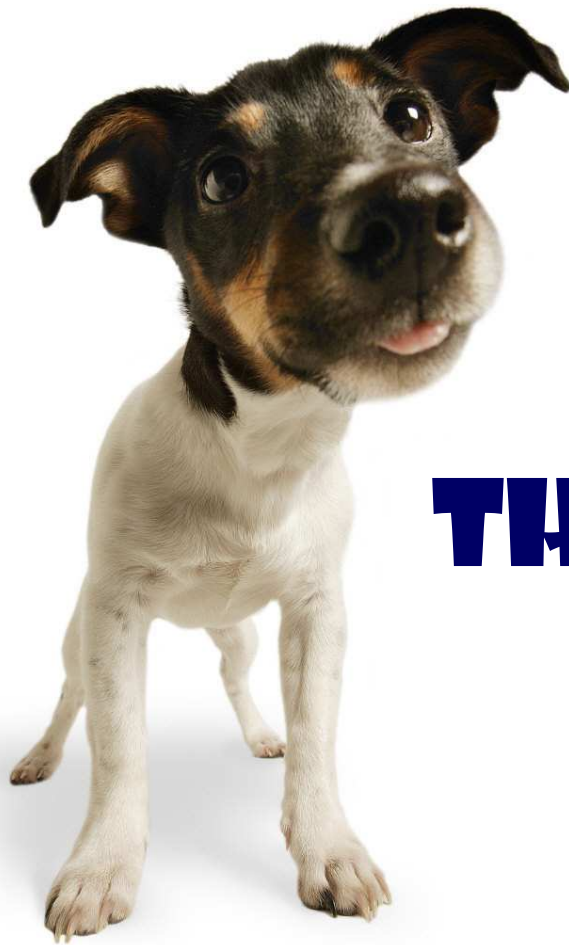
Practical maturity: in infancy, barely crawling.

Very hard problems remain to be tackled:

- operators with integrated assurances
 - confidentiality
 - privacy of access
 - correctness
- scalable protocols for secure hardware
 - massive data
 - good utilization of host CPUs
- areas
 - relational data
 - file systems
 - streaming data

/bin/yes > /dev/lunchtime

Stony Brook Network Security and Applied Cryptography Lab



THANK YOU !

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