



Computationally Secure Aggregation and PIR in the Shuffle Model

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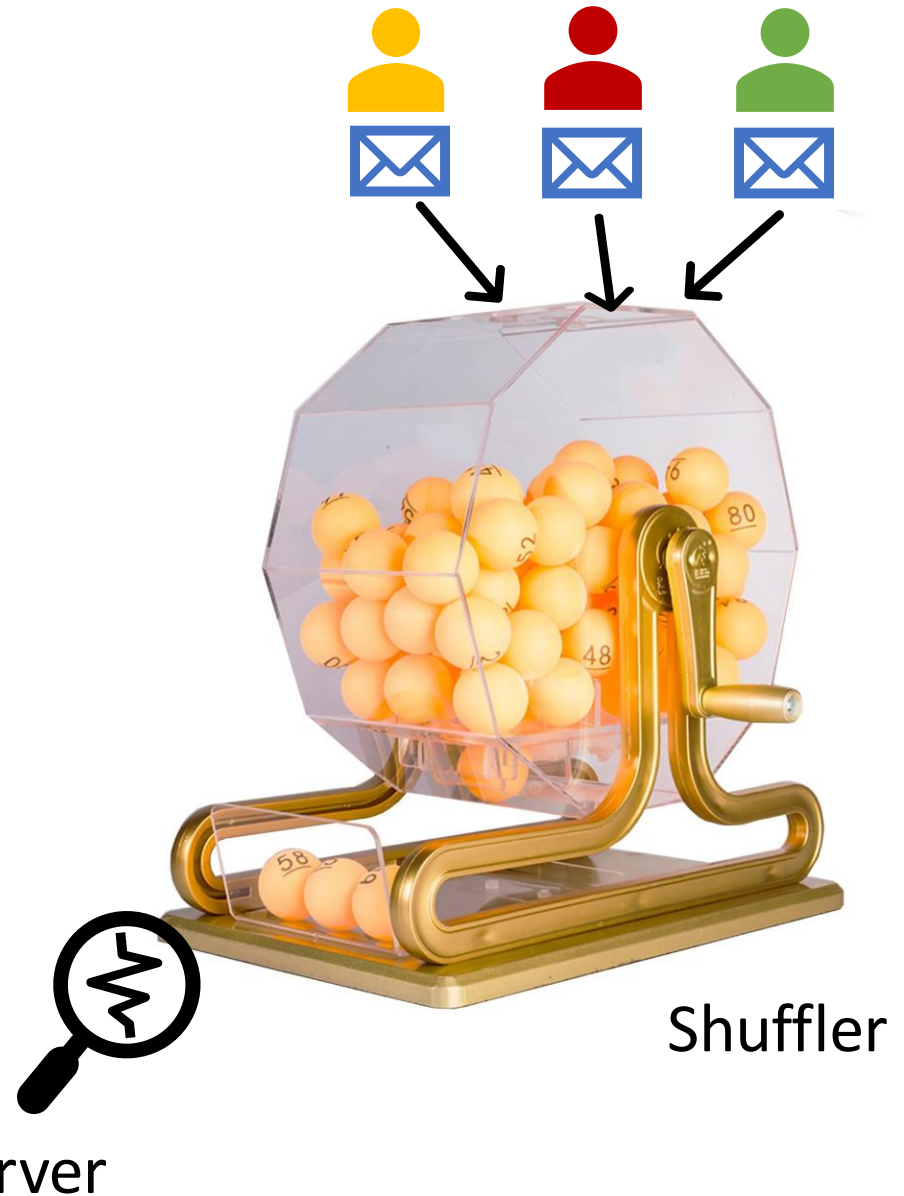


CORNELL
TECH



The shuffle model

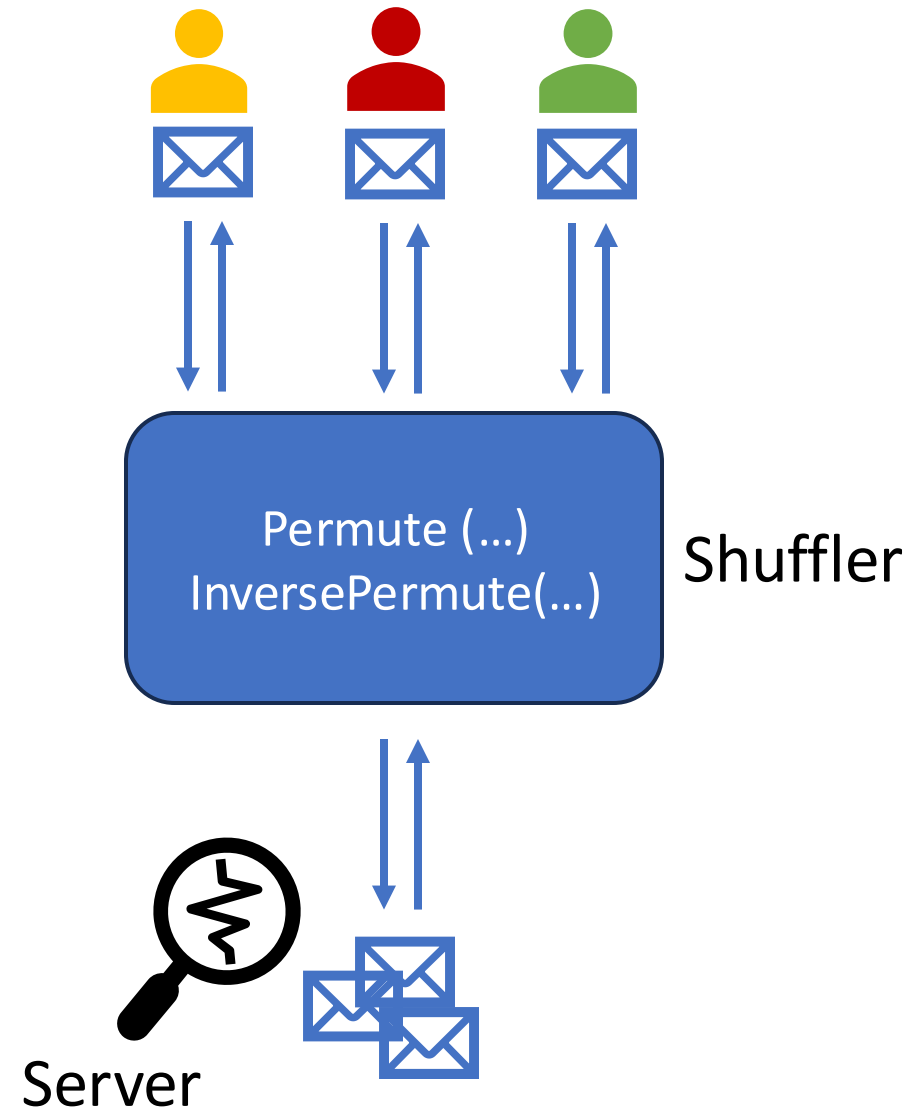
- Purpose: **anonymization**
- A popular model in differential privacy community
 - [Bittau et al. 2017]
 - [Cheu et al. 2019]
 - [Erlingsson et al. 2019]
 - ...
- Can be instantiated by, e.g., Tor



The shuffle model

- Purpose: **anonymization**
- A popular model in differential privacy community
- Can be instantiated by, e.g., Tor
- Later in our PIR setting:
 - We assume it is two-way
 - Can be viewed as a second shuffle server who does not hold the database

A hybrid model between single-server and two-server



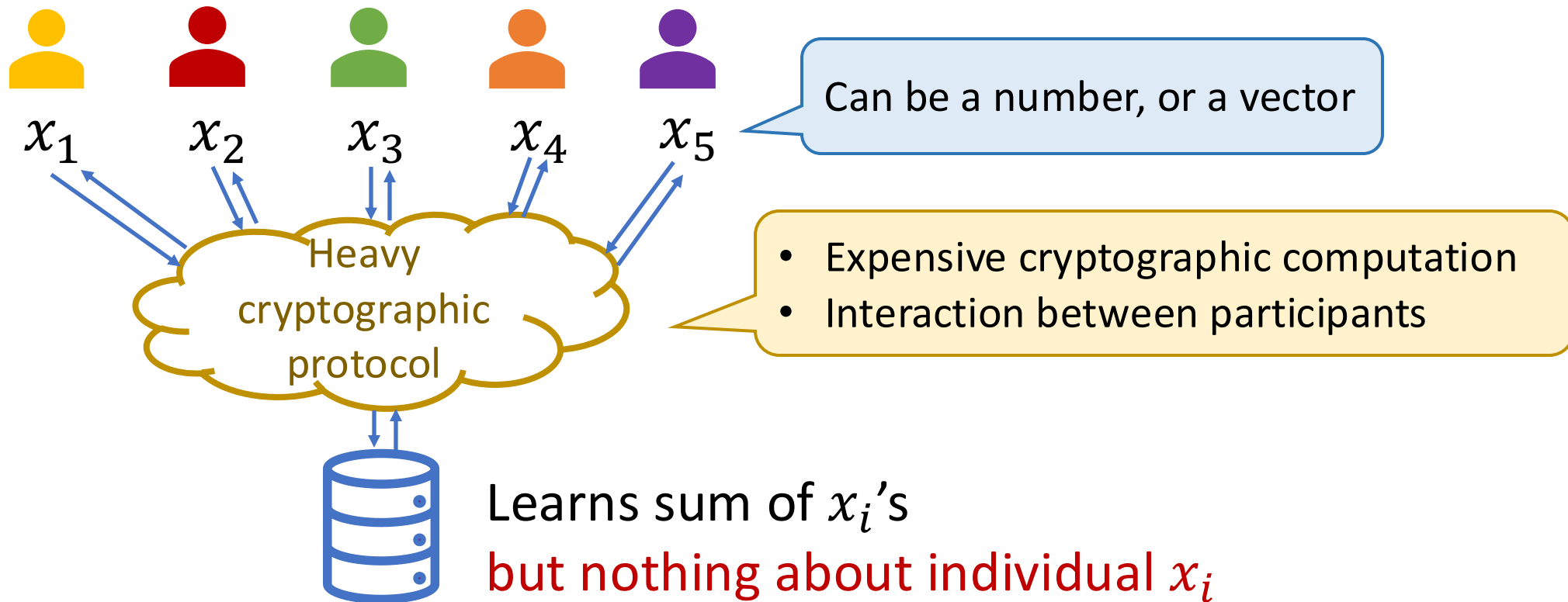
The main theme

Secure aggregation

Private information retrieval (PIR)

Improving efficiency of secure computation tasks
utilizing the shuffler

Single-server secure aggregation



Introducing anonymity into this problem



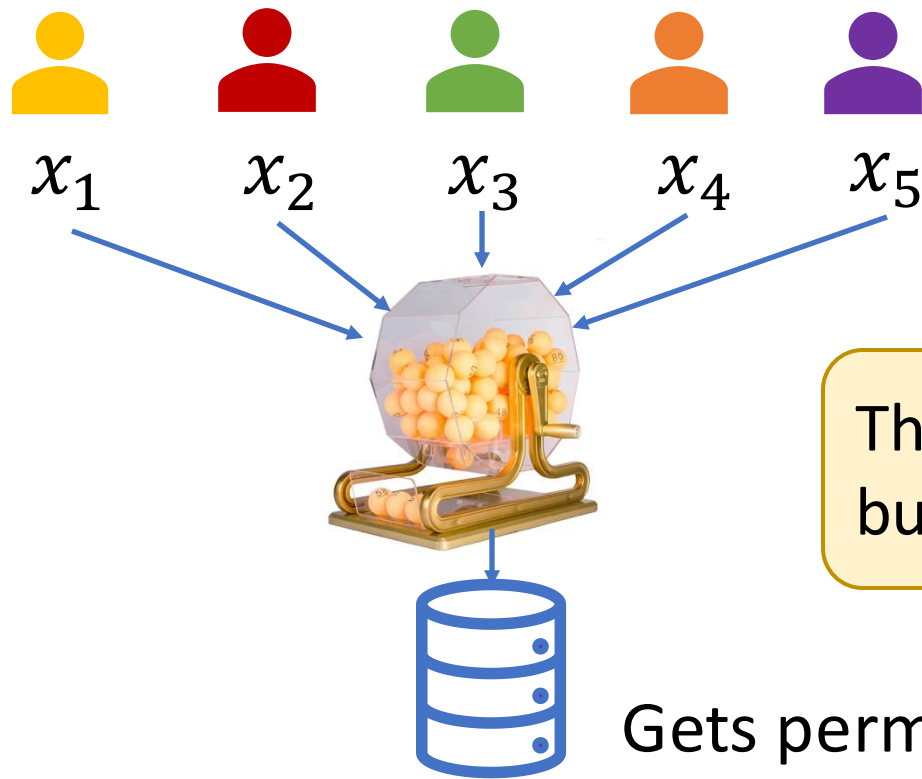
What we can get:

- Lightweight local computation
- Non-interactive



Learns sum of x_i 's
but nothing about individual x_i

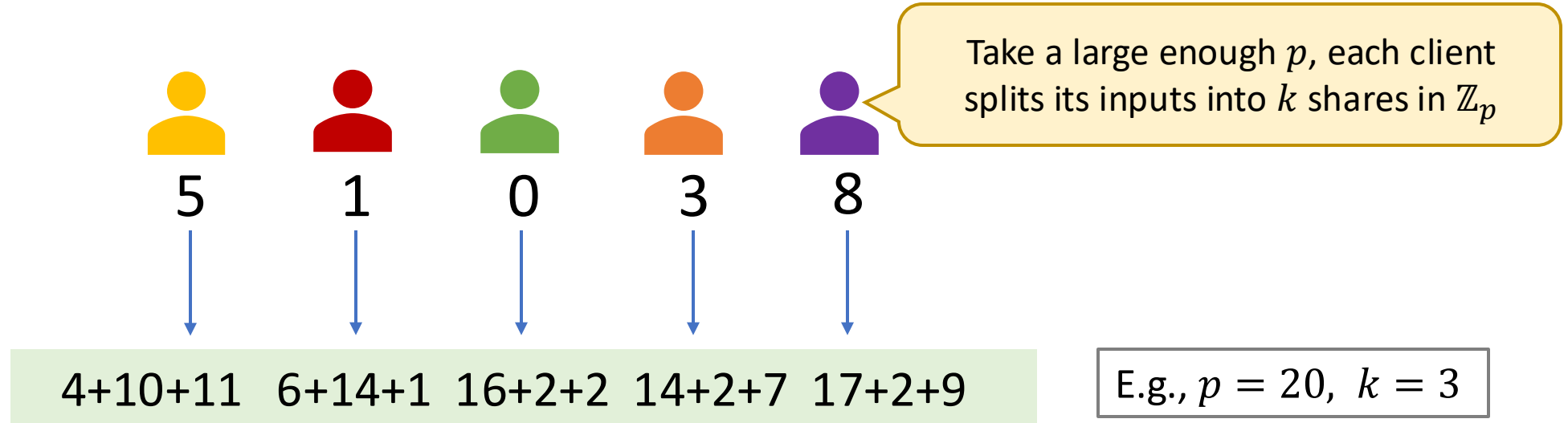
Anonymity does not trivialize the problem



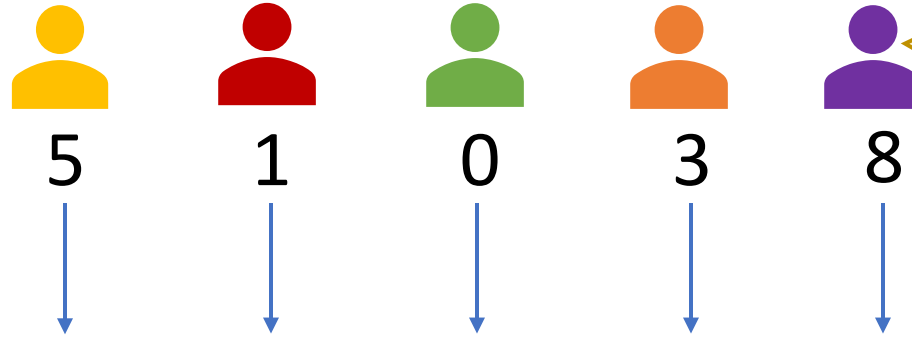
The shuffler hides who sends which message, but does not hide the message itself

Gets permuted x_i 's, adds them up

The split-and-mix paradigm [IKOS06]



The split-and-mix paradigm [IKOS06]



Take a large enough p , each client splits its inputs into k shares in \mathbb{Z}_p

4 10 11 6 14 1 16 2 2 14 2 7 17 2 9



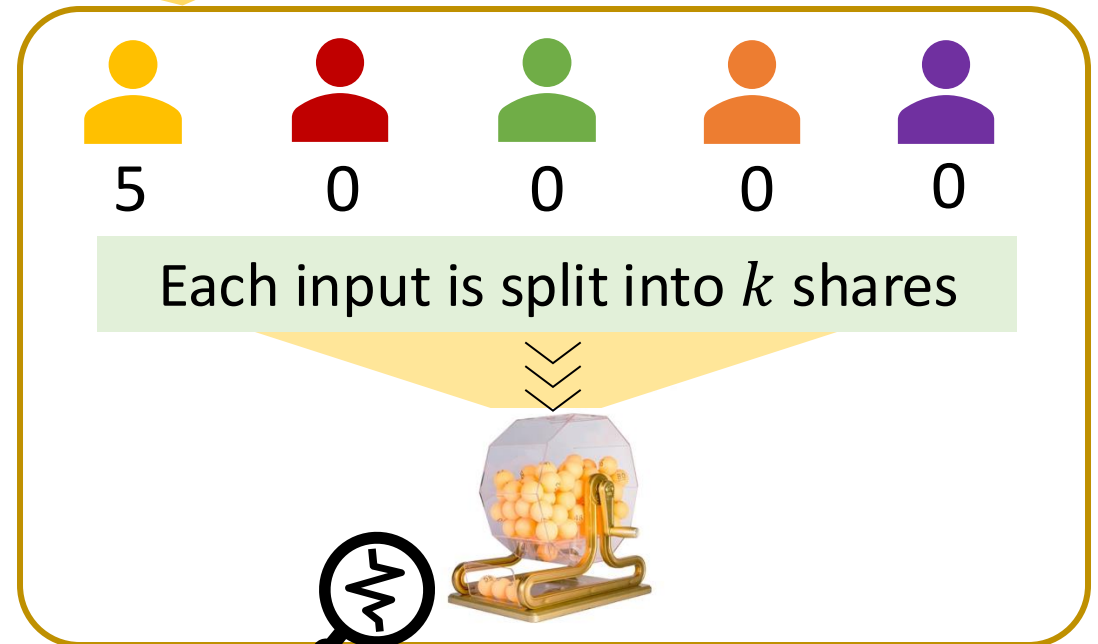
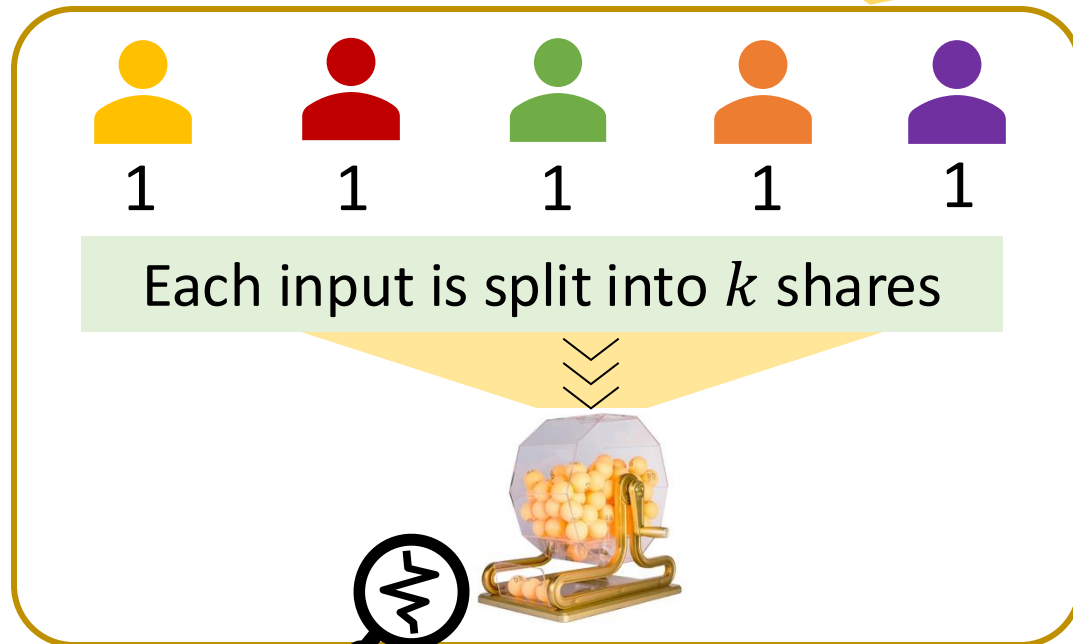
Shuffle all the shares

Sum up all the shares in \mathbb{Z}_p

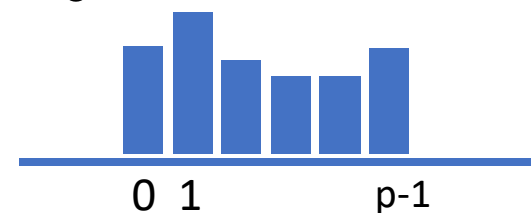
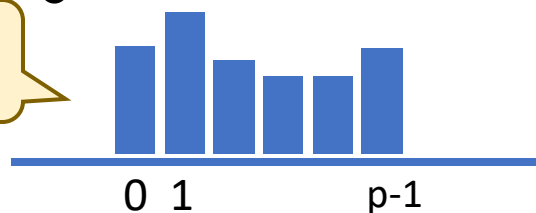


Security of split-and-mix

Any two different sets of inputs with equal sum

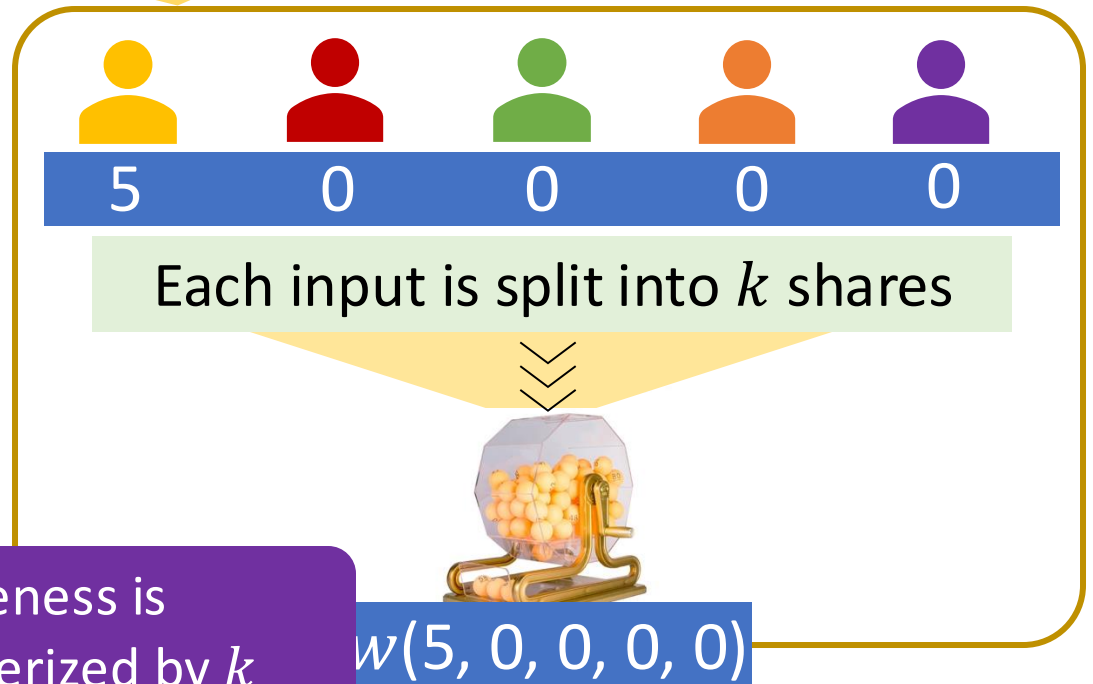
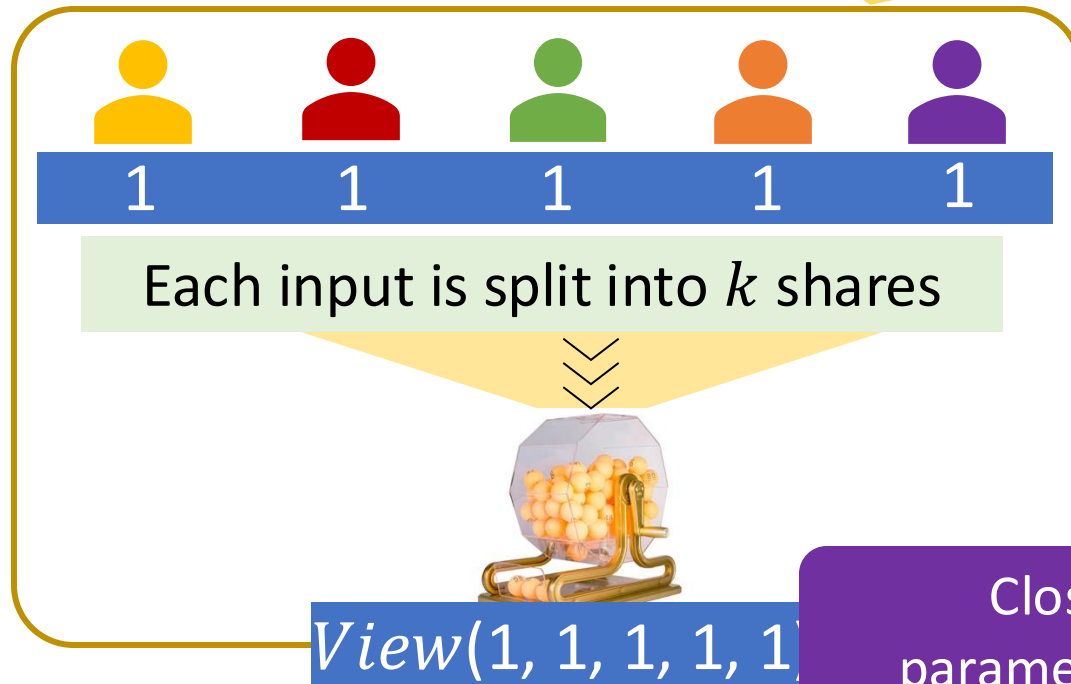


$View(1,1,1,1,1)$

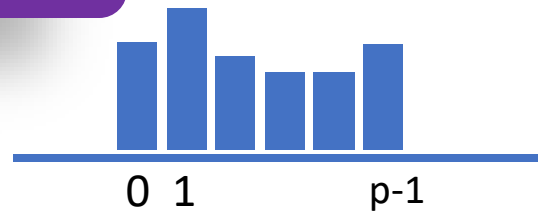
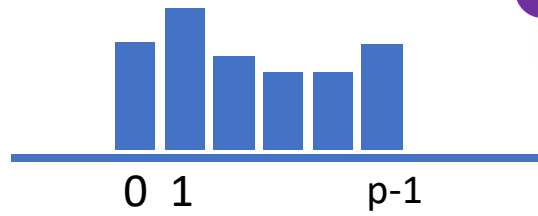


Security of split-and-mix

Any two different sets of inputs with equal sum



Closeness is parameterized by k



$$\text{View}(1, 1, 1, 1, 1) \approx \text{View}(5, 0, 0, 0, 0)$$

- Prior works only study statistical security [IKOS06, GMPV20, BBGN20]

#Clients	100	1000	10000
#Shares k (IT. 40 bits)	6317	3856	2775

Each client input: a vector $2^{15} \times \mathbb{F}_2$

New: computational security for split-and-mix

$$\text{View}(1, 1, 1, 1, 1) \approx \text{View}(5, 0, 0, 0, 0)$$

- Prior works only study statistical security [IKOS06, GMPV20, BBGN20]
- This work studies **computational security**, aiming to reduce the #shares k (and hence improving concrete efficiency)

#Clients	100	1000	10000
#Shares k (IT. 40 bits)	6317	3856	2775
#Shares k (Comp. 128 bits)	405	88	37

Each client input: a vector $2^{15} \times \mathbb{F}_2$

Our results

Computational security for split-and-mix based on SD, MDSD

```
graph TD; A[Computational security for split-and-mix based on SD, MDSD] --> B[Single-server secure aggregation in the shuffle model]; A --> C[Single-server PIR in the shuffle model];
```

Single-server secure aggregation
in the shuffle model

Up to 25X savings for communication
compared to the best statistical split-
and-mix baseline

(Even giving advantage to the baseline
by compressing the shares)

Single-server PIR
in the shuffle model

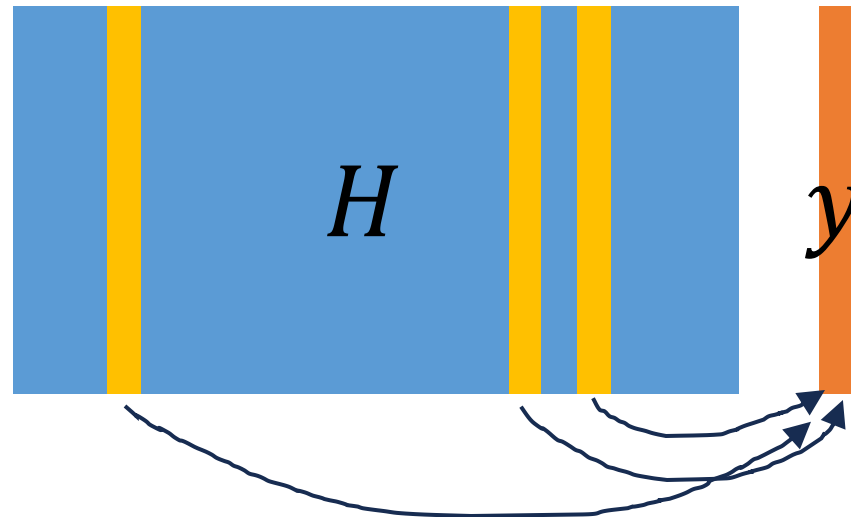
Up to 22X improvement of throughput (in the
batch setting) over SimplePIR [HHCMV23] with
comparable communication cost

Split-and-mix based on Syndrome Decoding (SD)

- The SD assumption (dual-LPN [BFKL94, AIK07])
 H : a random matrix
 y : a target vector (e.g., a client's input)



Computationally hard to find low-weight vector e such that $H \cdot e = y$



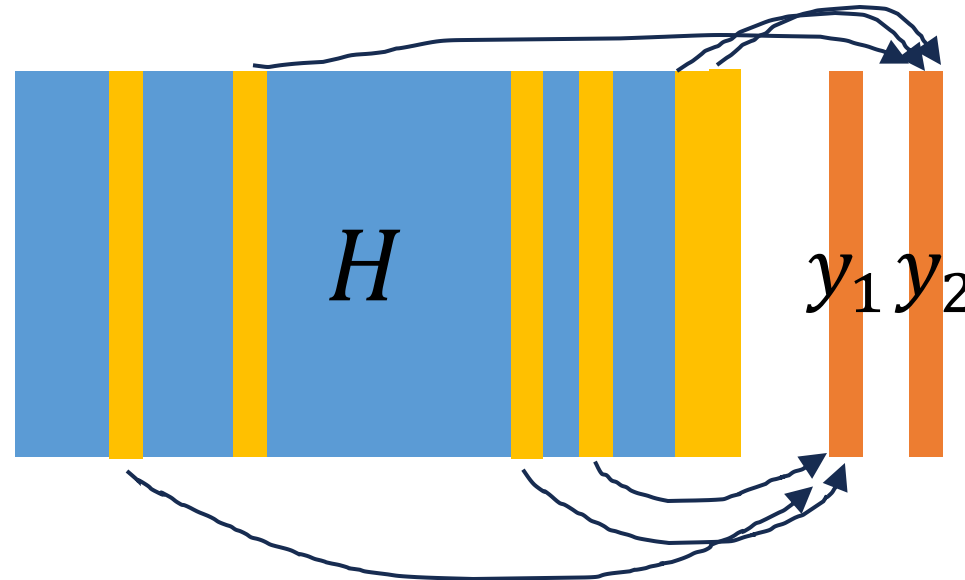
Split-and-mix based on Syndrome Decoding (SD)

- “Multi-Disjoint” Syndrome Decoding
 H : a random matrix
 $Y = [y_1, y_2, \dots]$: multiple target vectors (e.g., multiple client inputs)

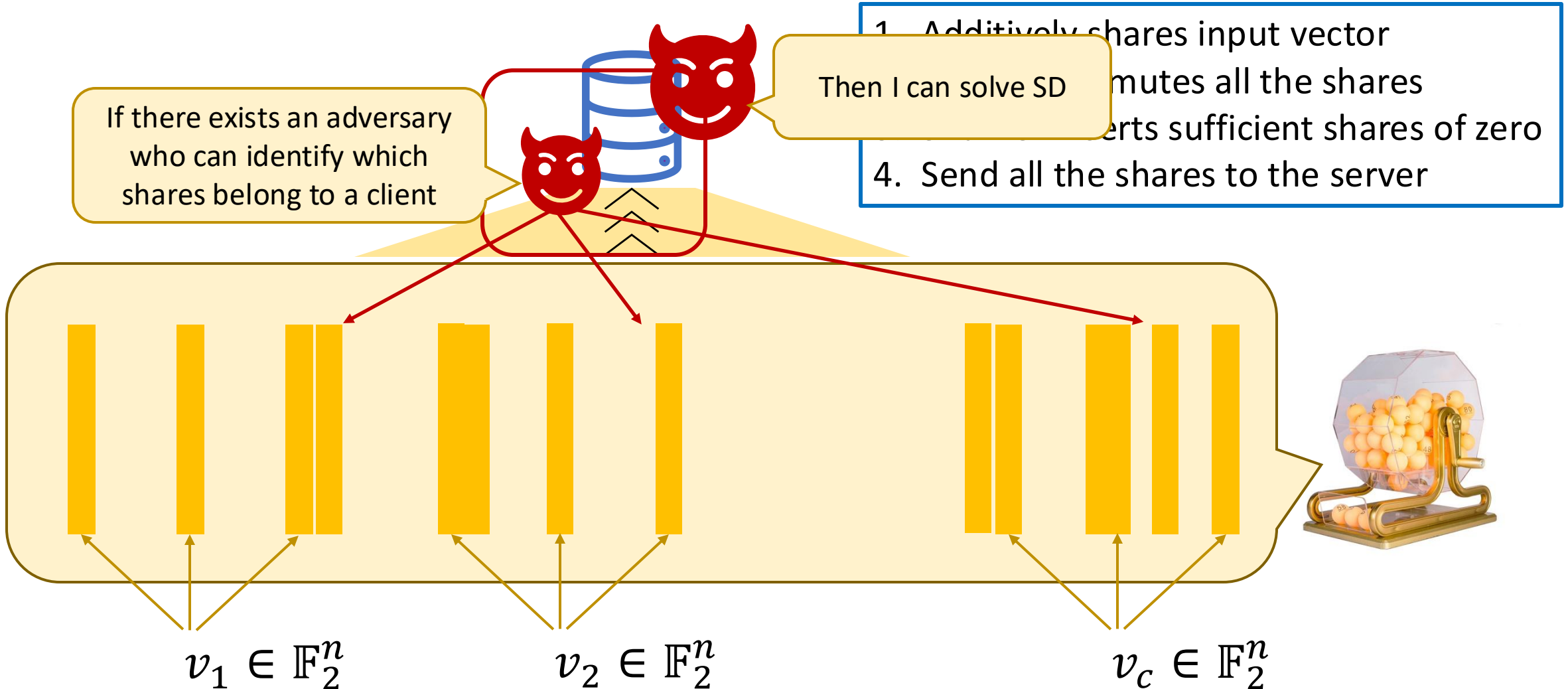


Computationally hard to find “low-weight” E such that $H \cdot E = Y$

We generalize SD to
Multi-Disjoint Syndrome Decoding
to handle multiple clients



The resulting aggregation protocol in a nutshell



Our results

Computational security for split-and-mix based on LPN, MDSD



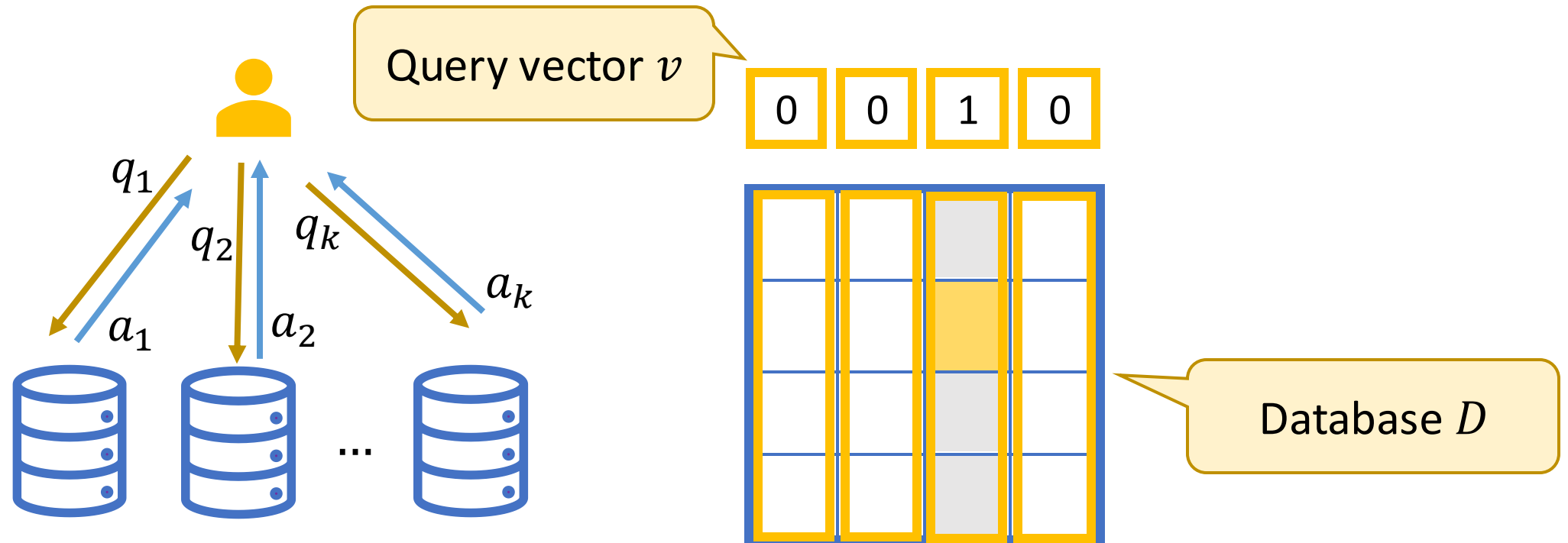
Single-server secure aggregation
in the shuffle model

Up to 25X savings for communication
compared to the best baseline in the
statistical setting

Single-server PIR
in the shuffle model

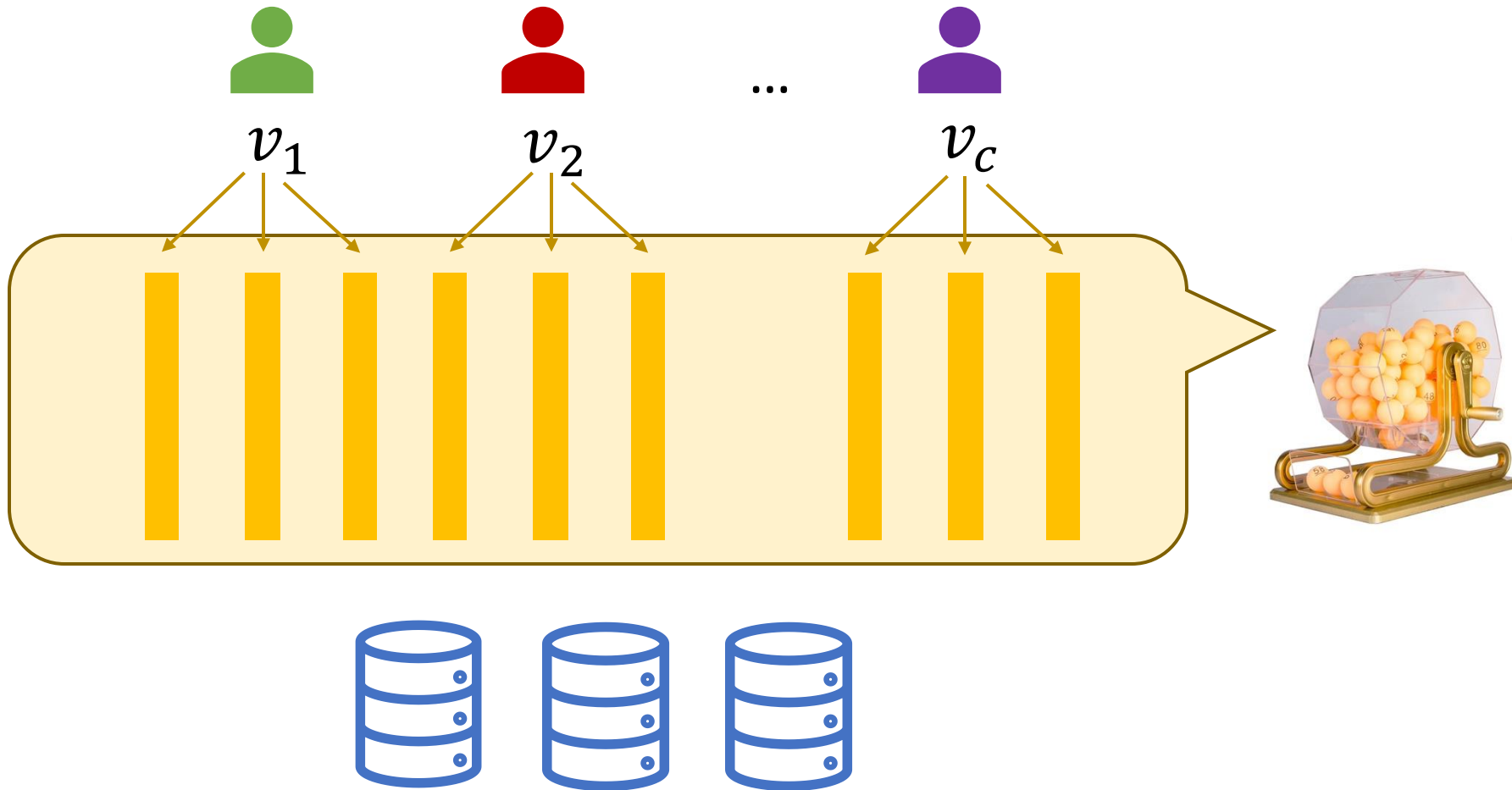
Up to 22X improvement of throughput (in the
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Starting point: a classic multi-server PIR

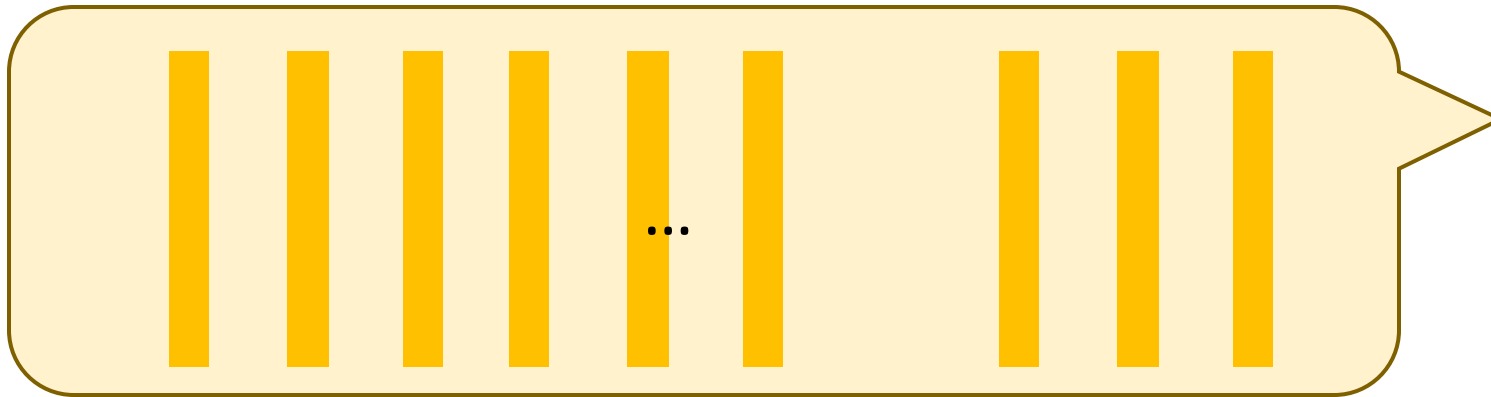


$$\begin{aligned} v \cdot D &= (q_1 + \dots + q_k) \cdot D \\ &= q_1 \cdot D + \dots + q_k \cdot D \\ &:= a_1 + \dots + a_k \end{aligned}$$

Single-server PIR from split-and-mix



Single-server PIR from split-and-mix



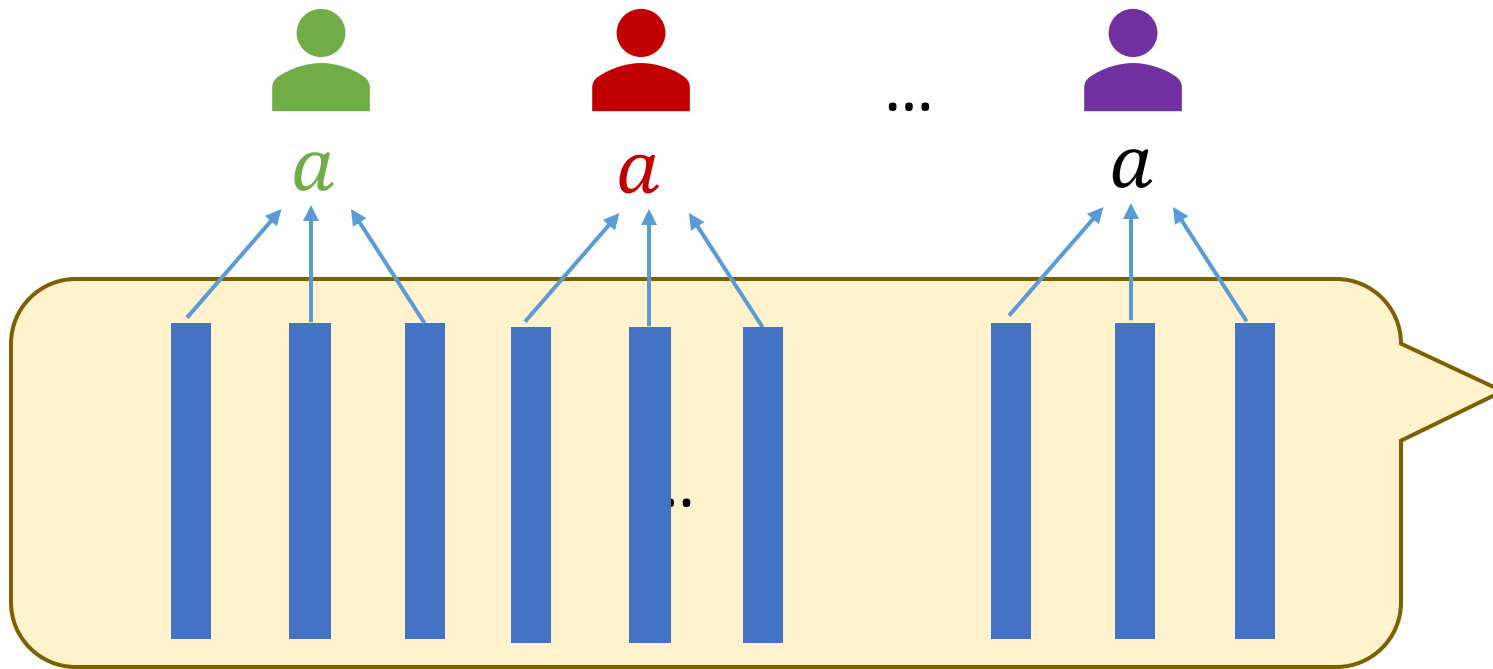
Two-way
anonymous channel



A blue icon of a database cylinder with three horizontal bands and small circles representing data points.
$$\langle D, \text{yellow bar} \rangle = \text{blue bar}$$

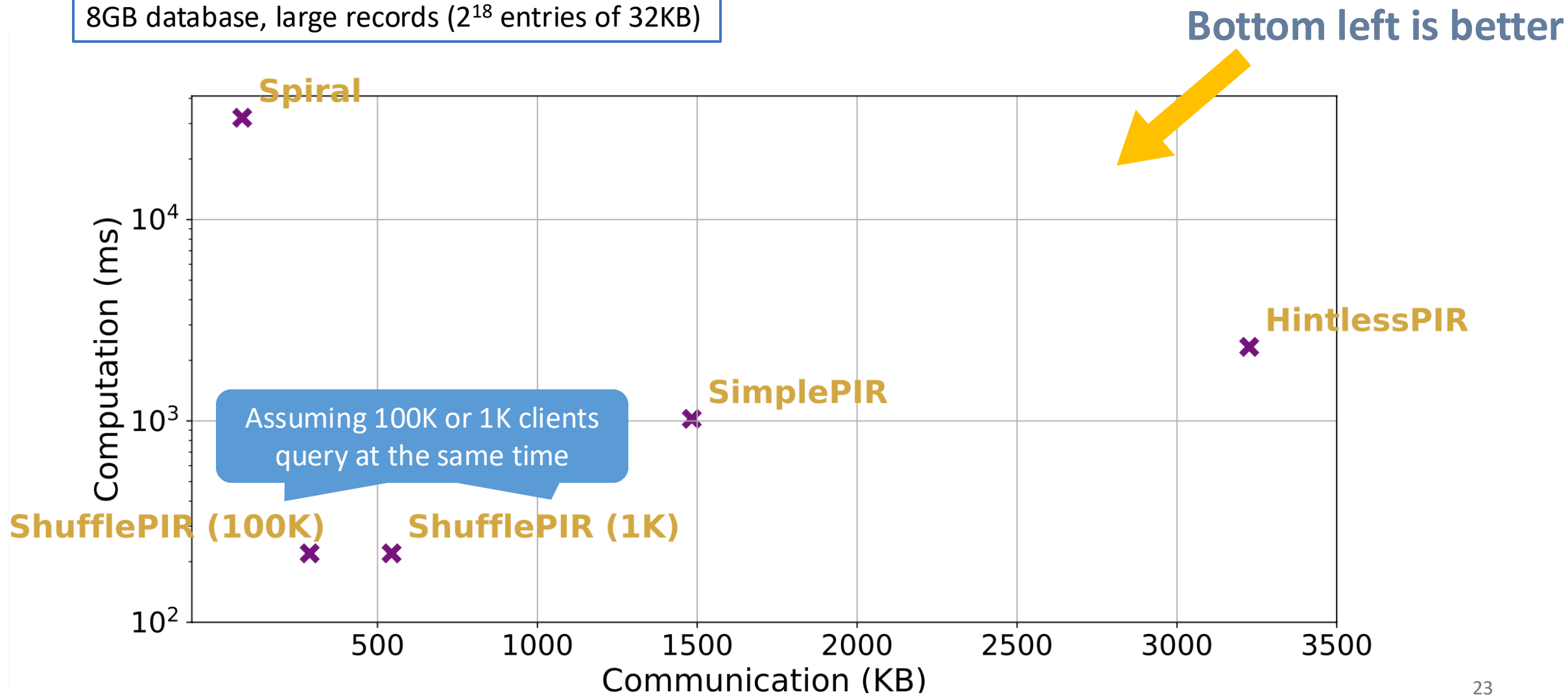
Single-server PIR from split-and-mix

[IKOS06] initialized the study of PIR from split-and-mix, but their construction is rather theoretical



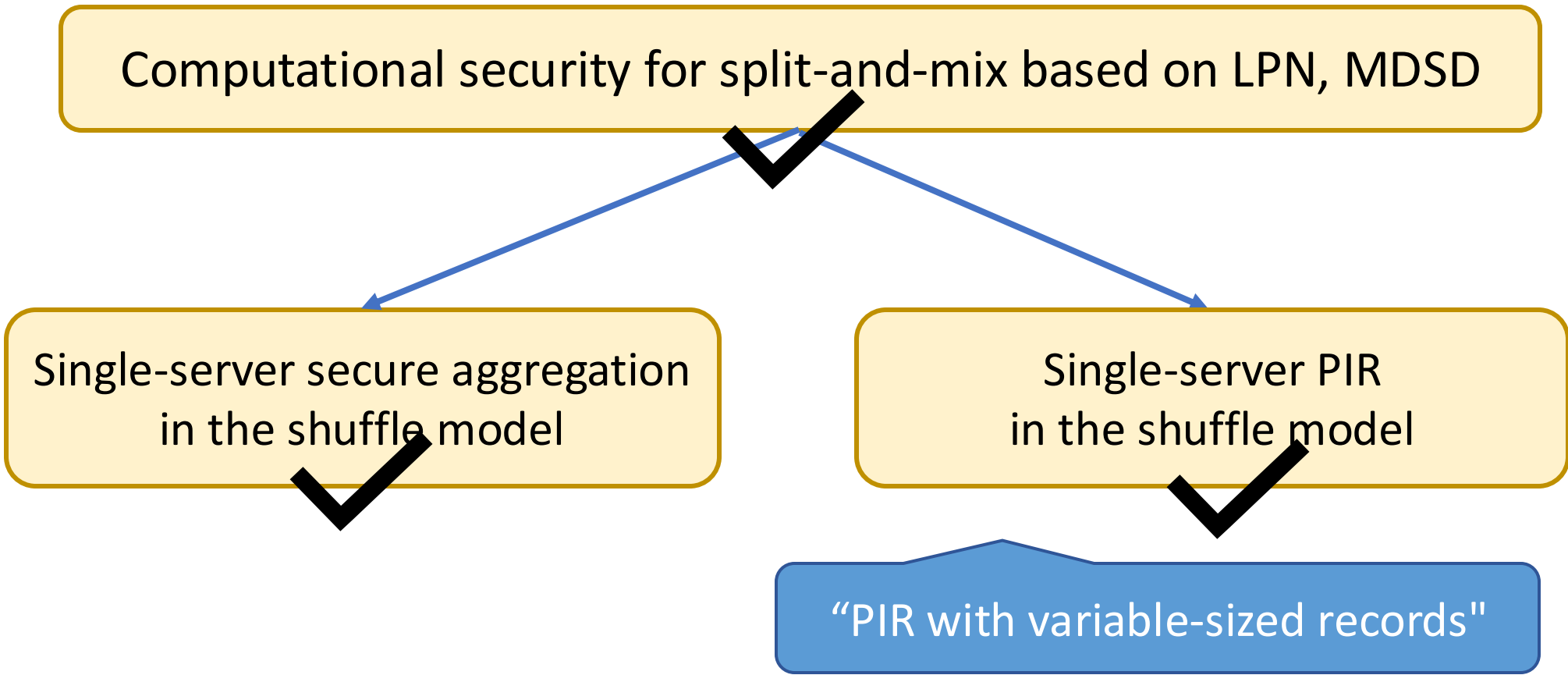
Performance

8GB database, large records (2^{18} entries of 32KB)



Thanks!

Summary



Backup slides

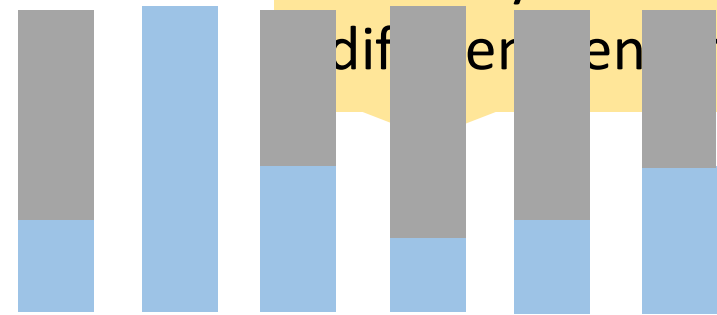
PIR with variable-sized records

- Deploying PIR in real-world applications

Often assume the same length



Database entries of PIR *in theory*



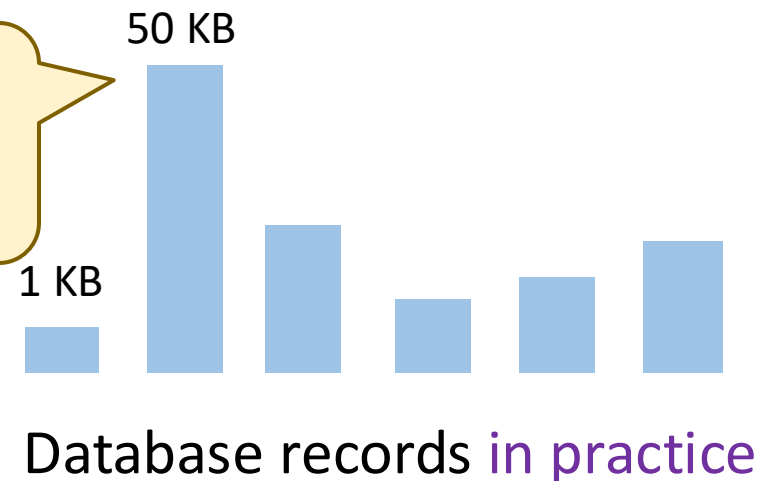
Database records *in practice*

To retrieve privately, it is necessary to hide record size

PIR with variable-sized records

- Padding solves the problem, but it is inefficient for some applications

Client who retrieves the small record has to pay the cost of retrieving the largest record



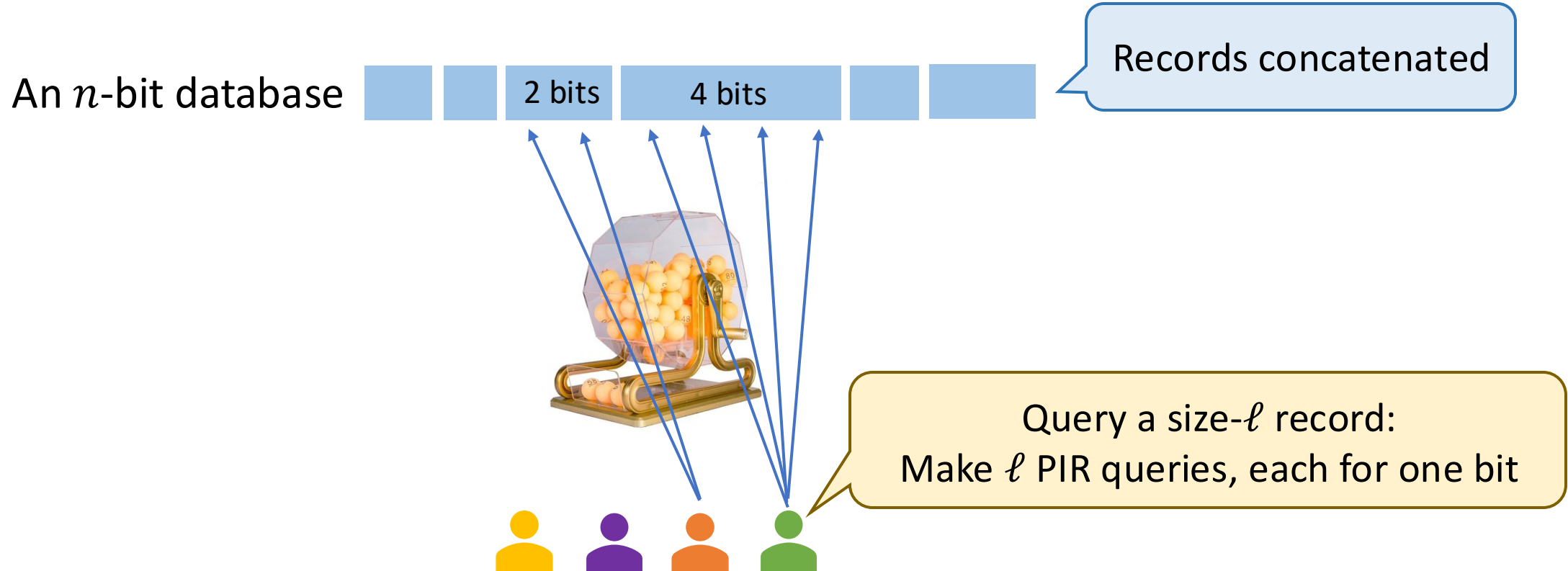
To retrieve privately, it is necessary to hide record size

PIR with variable-sized records

- In the standard model, there is no way out
- In the shuffle model:
 - Client communication proportional to the length of the retrieved record
 - Leak only the total size of all queried records (in most cases quite benign)

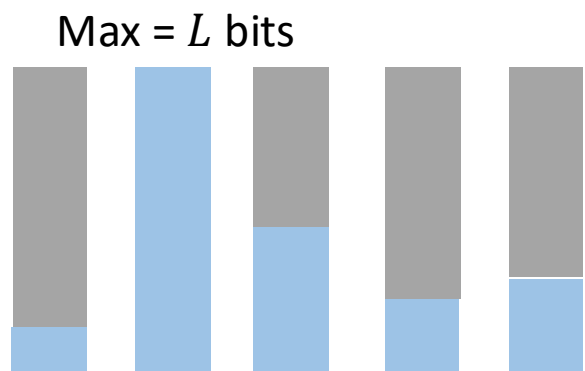
PIR with variable-sized records

- A toy protocol using PIR as a black box



PIR with variable-sized records

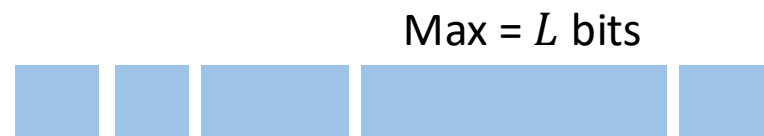
To retrieve an ℓ -bit record (out of T records of total n bits)



1 PIR query on DB
of size $LT \gg n$

No leakage on
queried record length

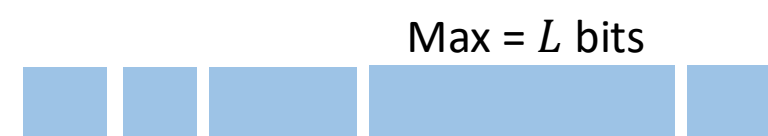
Standard model



Polylog ℓ PIR queries
on DB of size n

Leak total length of
queried records + a bit more

Shuffle model, our construction



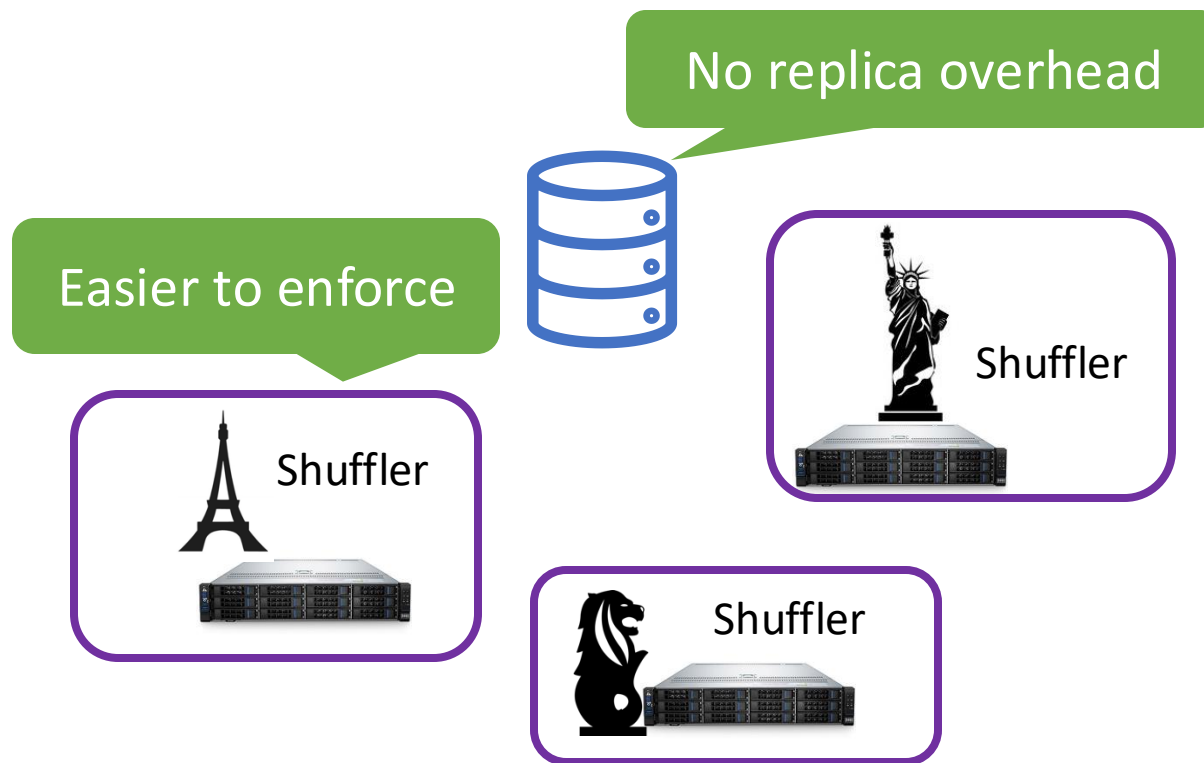
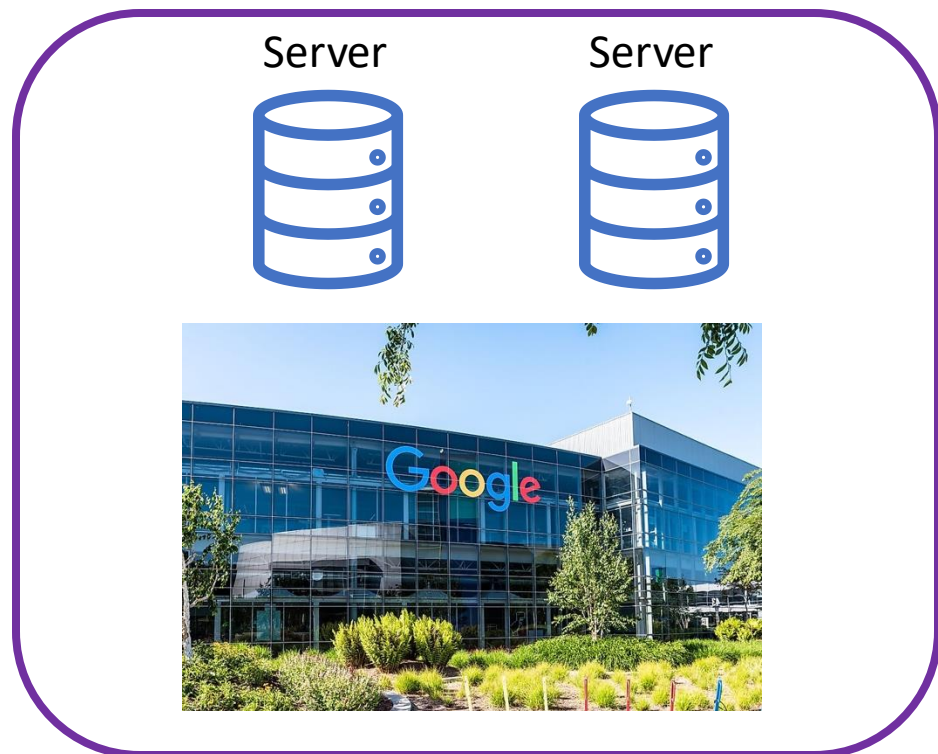
ℓ PIR queries
on DB of size n

Leak total length of
queried records

Shuffle model, toy version

Discussion

- Assuming non-colluding servers vs. assuming a two-way anonymous channel



Discussion

- Exploiting tradeoffs when designing protocols: making assumptions, relaxing security, etc.
- Guaranteeing different assumptions does not require the same amount of efforts: system efforts, law efforts, etc.
- The likelihood of assumptions being compromised in real-world scenarios may vary