

CS 350S: Privacy-Preserving Systems

Private aggregate statistics

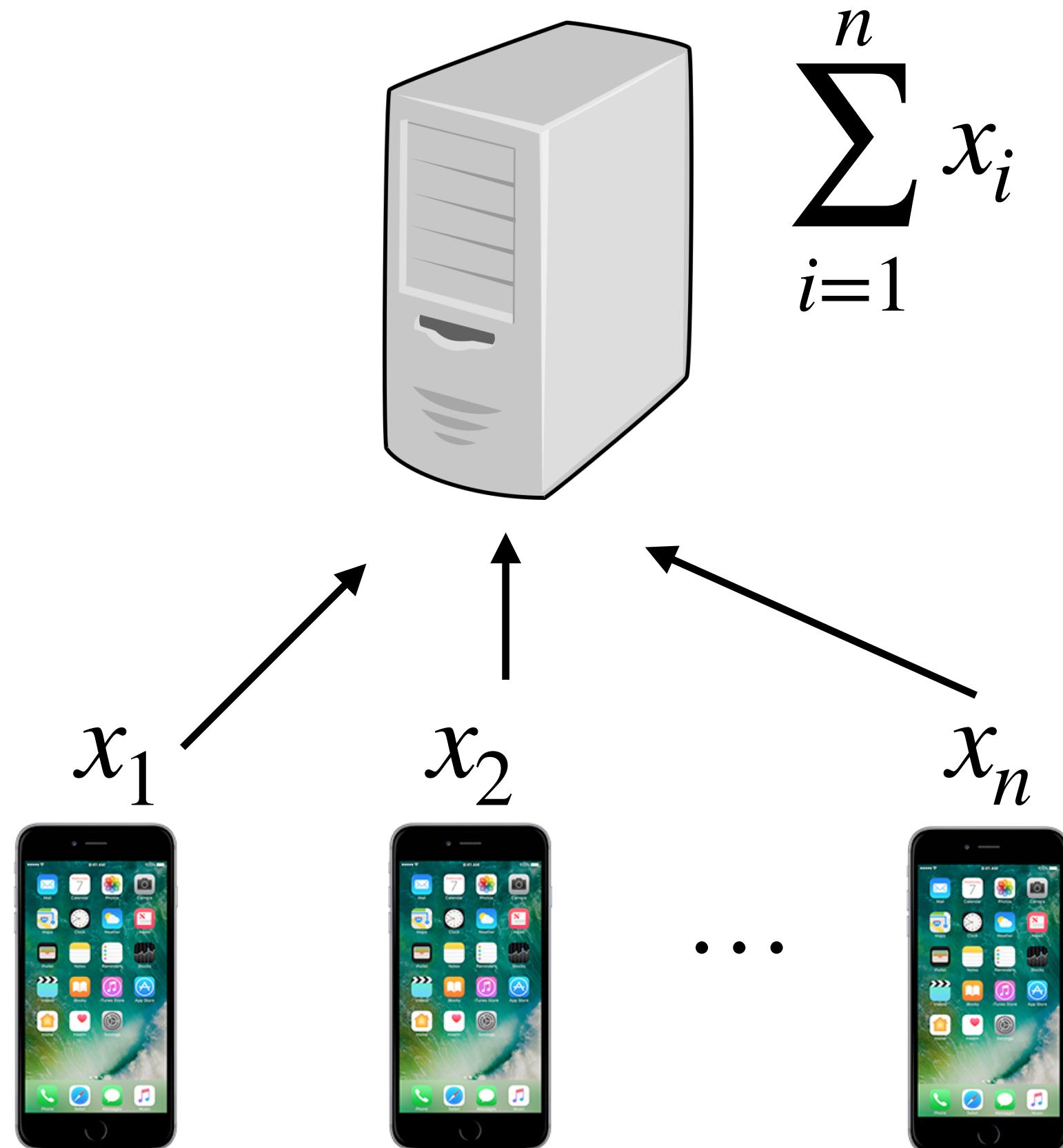
Outline

- 1. Prio design**
2. Tim Geoghegan and Chris Patton

Private aggregation

Mozilla wants to know:

- How many users disable content blocking on nytimes.com?

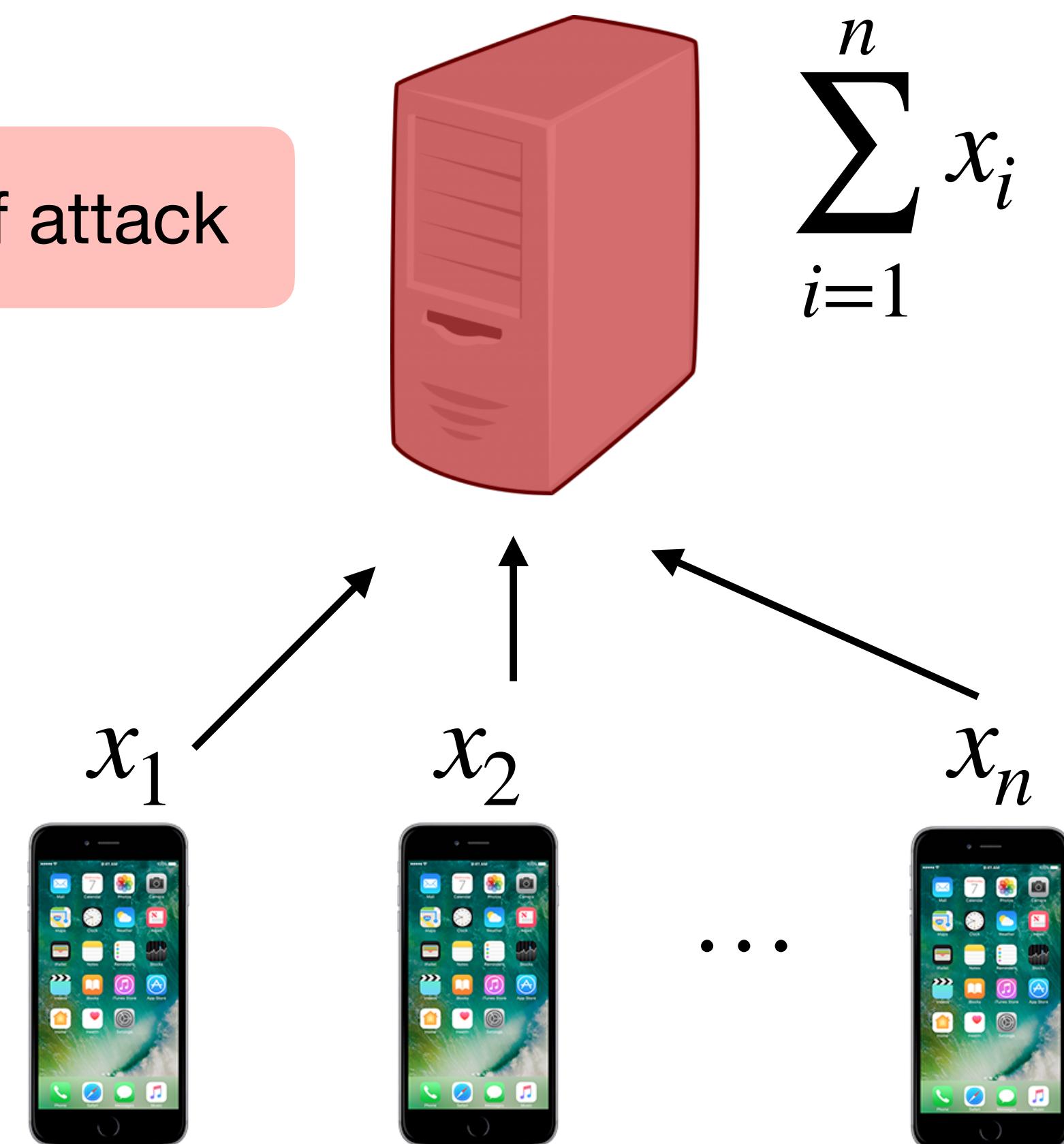


Private aggregation

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Server is central point of attack



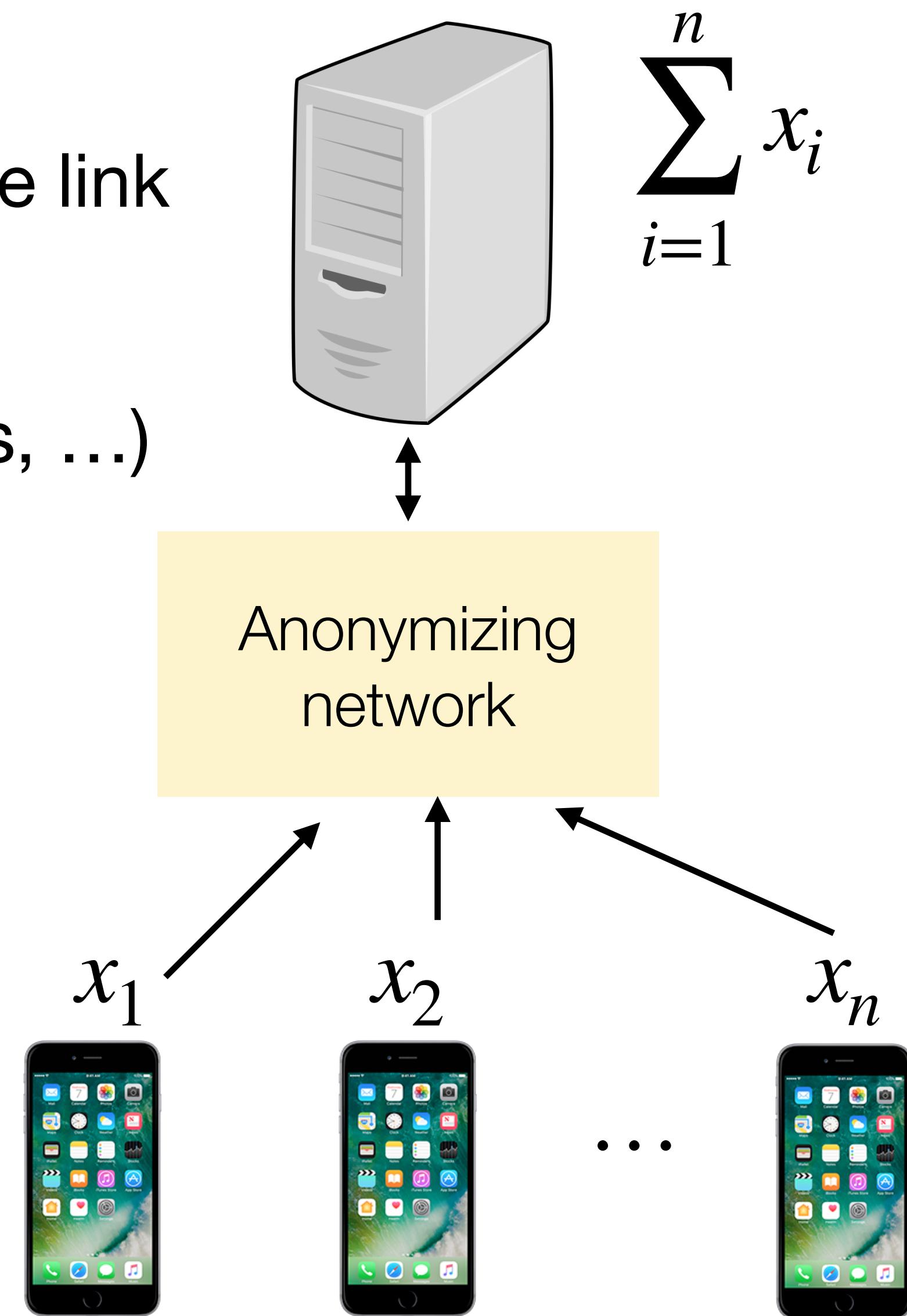
One approach

What tool from class could we use to hide the link between users and data submissions?

Anonymizing network (Tor, mix-nets, DC-nets, ...)

Drawbacks?

- Systems either difficult to scale or vulnerable to traffic-analysis attacks



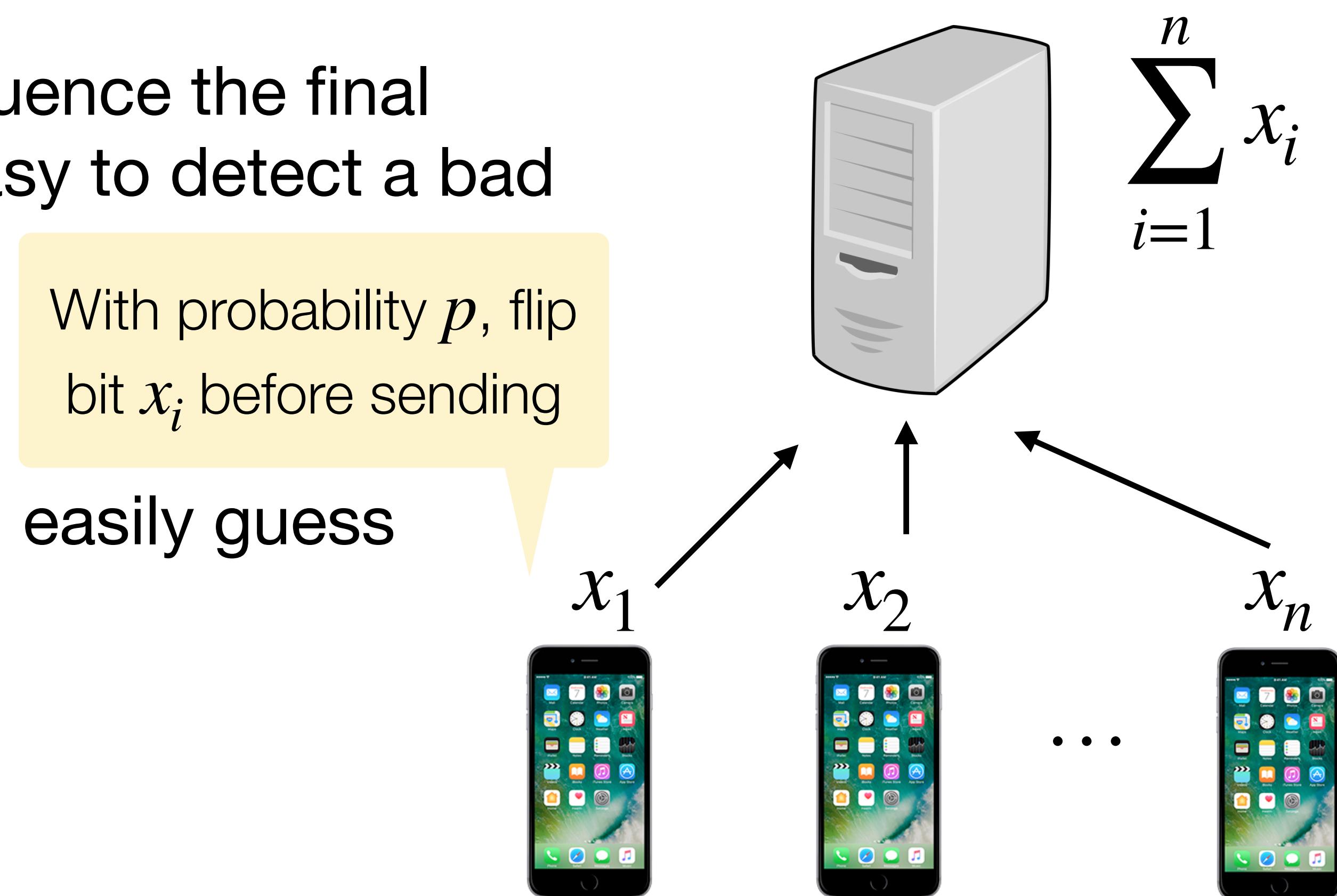
Another approach: randomized response

Advantages?

- Simple and scales well
- A malicious client can influence the final score by at most $+\text{-} 1$: easy to detect a bad score

Disadvantages?

- Small p : weak privacy (can easily guess original input)
- Large p : not very useful



One more approach: additively homomorphic encryption

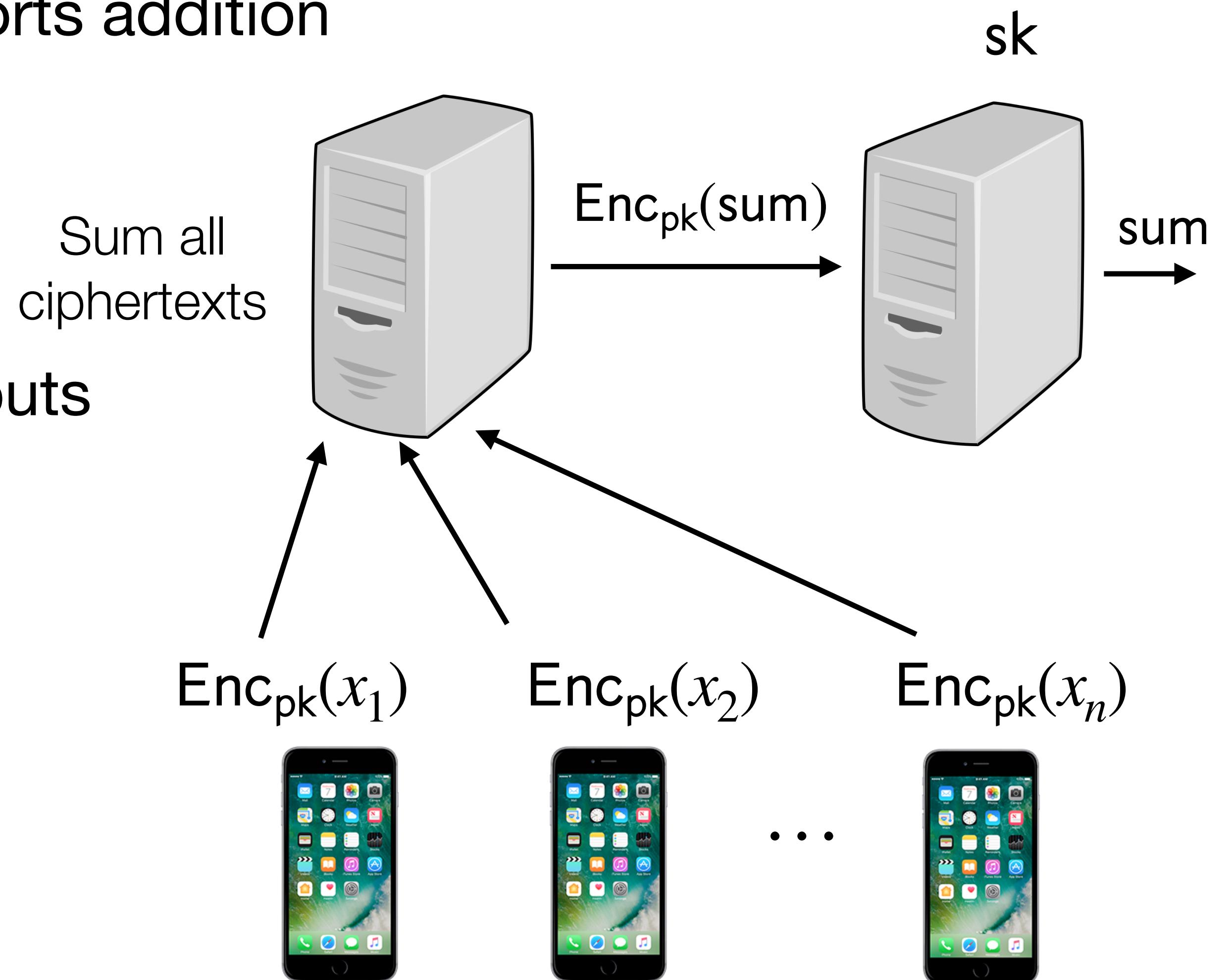
Tool: encryption scheme that supports addition

Advantages?

- Hides client contributions
- Correctly computes sum over inputs

Disadvantages?

- No protection against malicious client contributions
- No protection against malicious aggregator



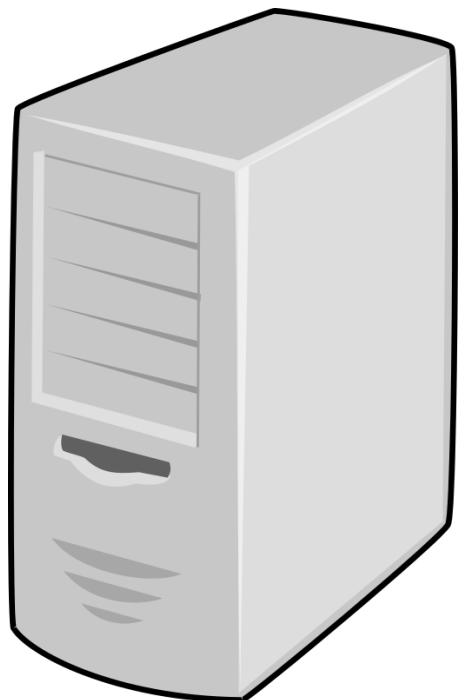
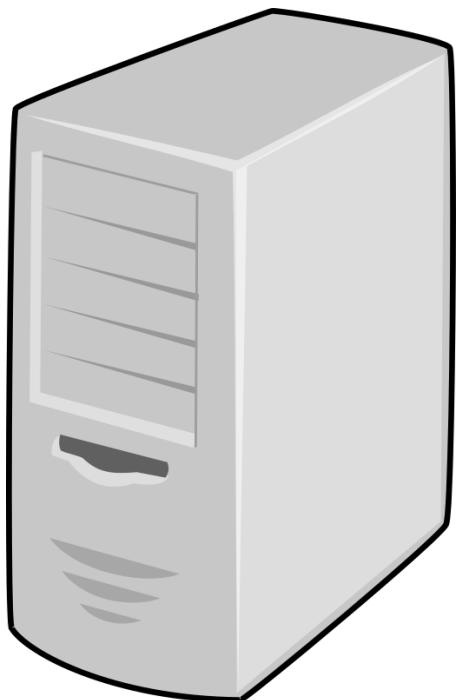
Prio

[Corrigan-Gibbs, Boneh]

All three of:

- Utility: Correctly computes sum
- Privacy: Hides user inputs
- Robustness: Malicious clients cannot overly influence sum

$$\sum_{i=1}^n x_i$$



x_1



x_2



\dots

x_n

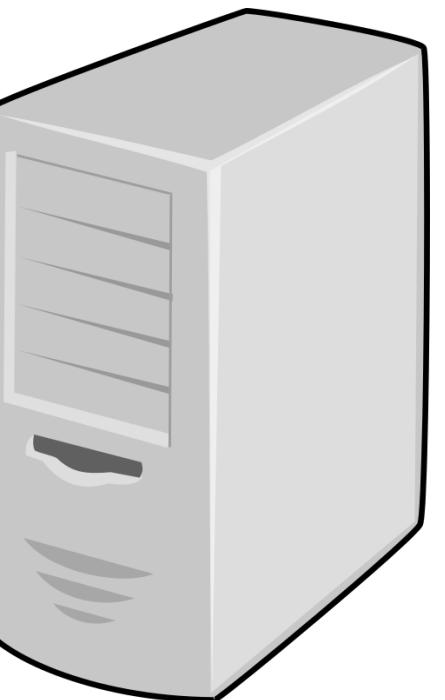


Prio properties

Aggregation function f

$$f(x_1, x_2, \dots, x_n)$$

Correctness: If all servers are honest, servers learn $f(x_1, x_2, \dots, x_n)$



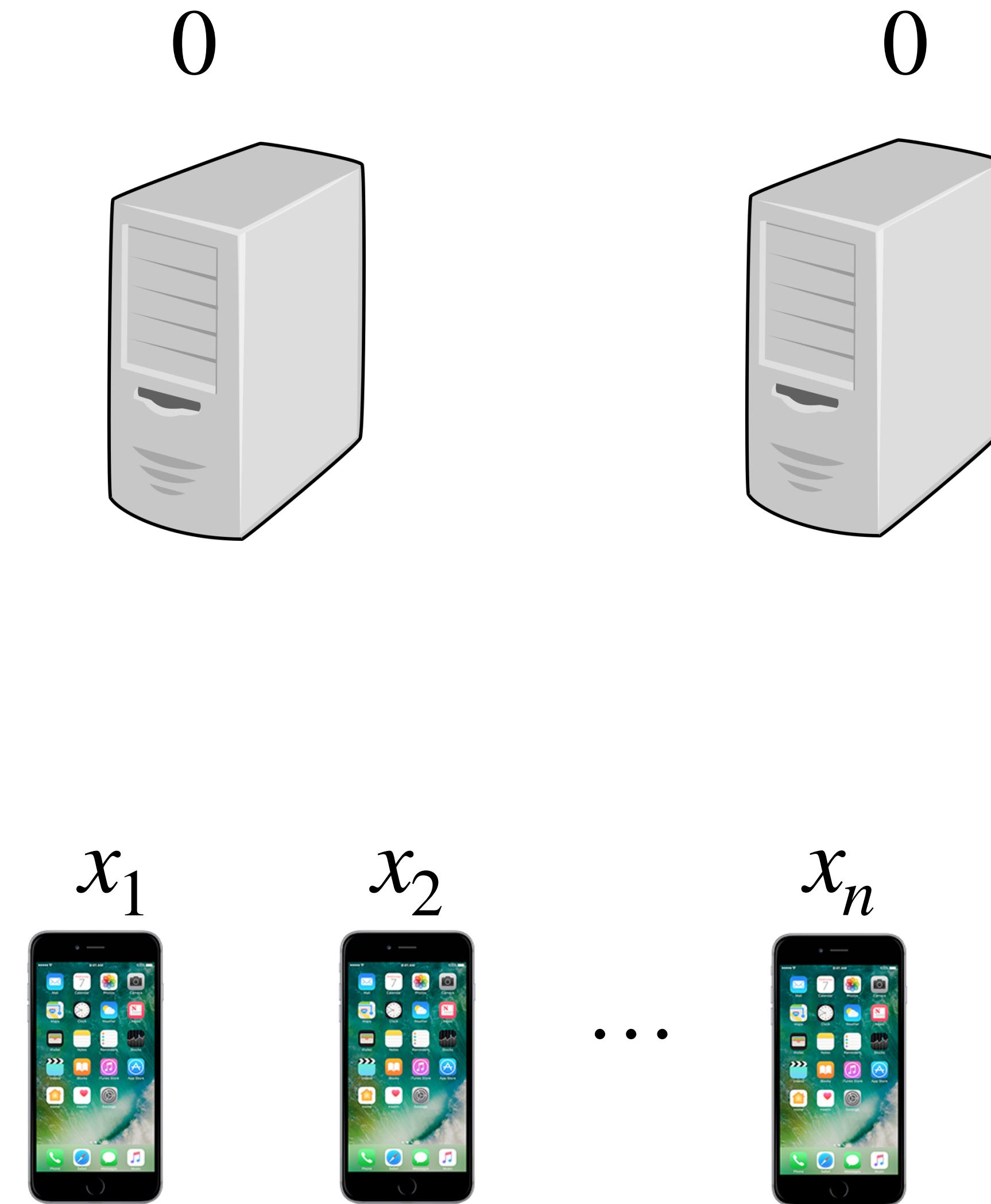
Privacy: If one server is honest, servers only learn $f(x_1, x_2, \dots, x_n)$

- Privacy with 1 malicious server

Robustness: Malicious clients have bounded influence



Warm up: computing private sums

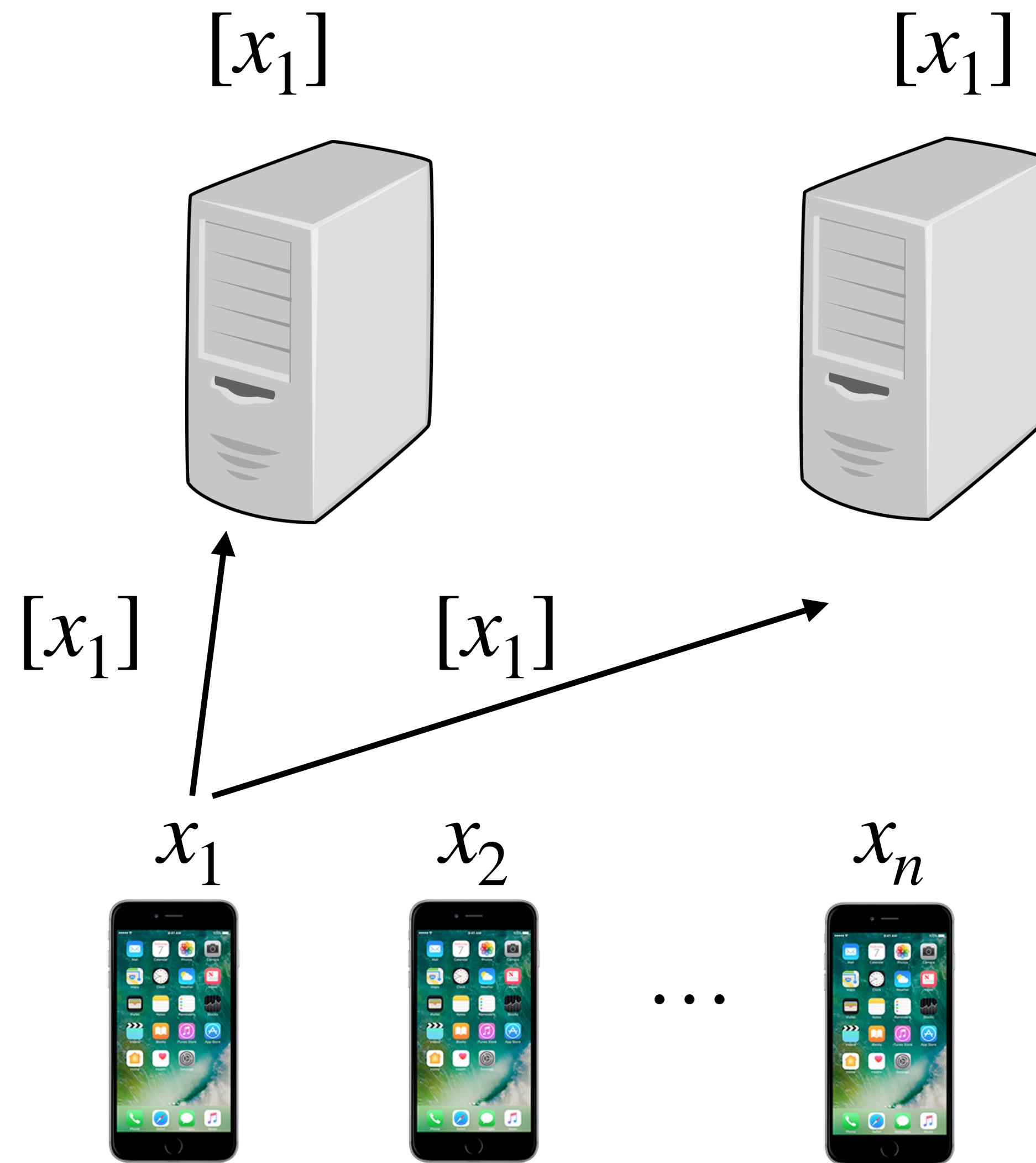


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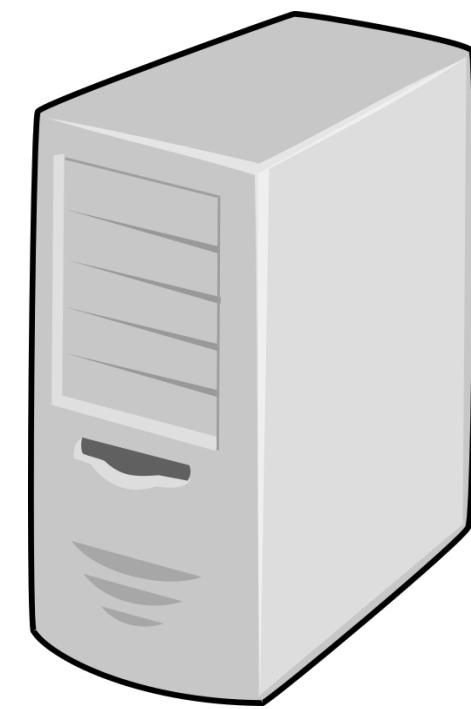
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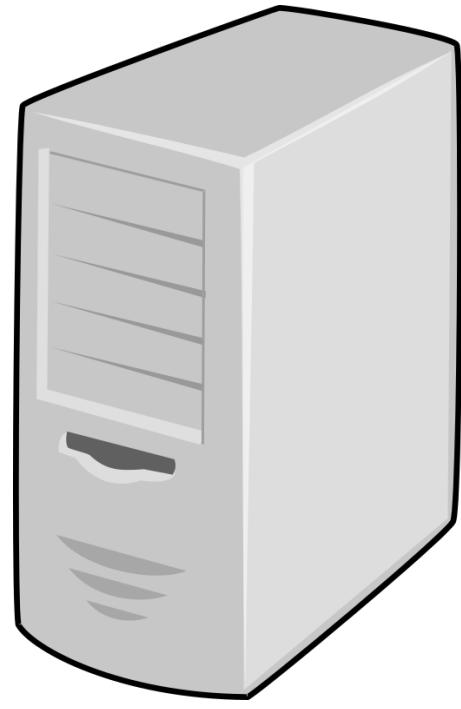
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Warm up: computing private sums

$$[x_1] + [x_2]$$



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$$[x_2]$$

$$x_1$$



$$[x_2]$$

$$x_2$$



$$x_n$$



...

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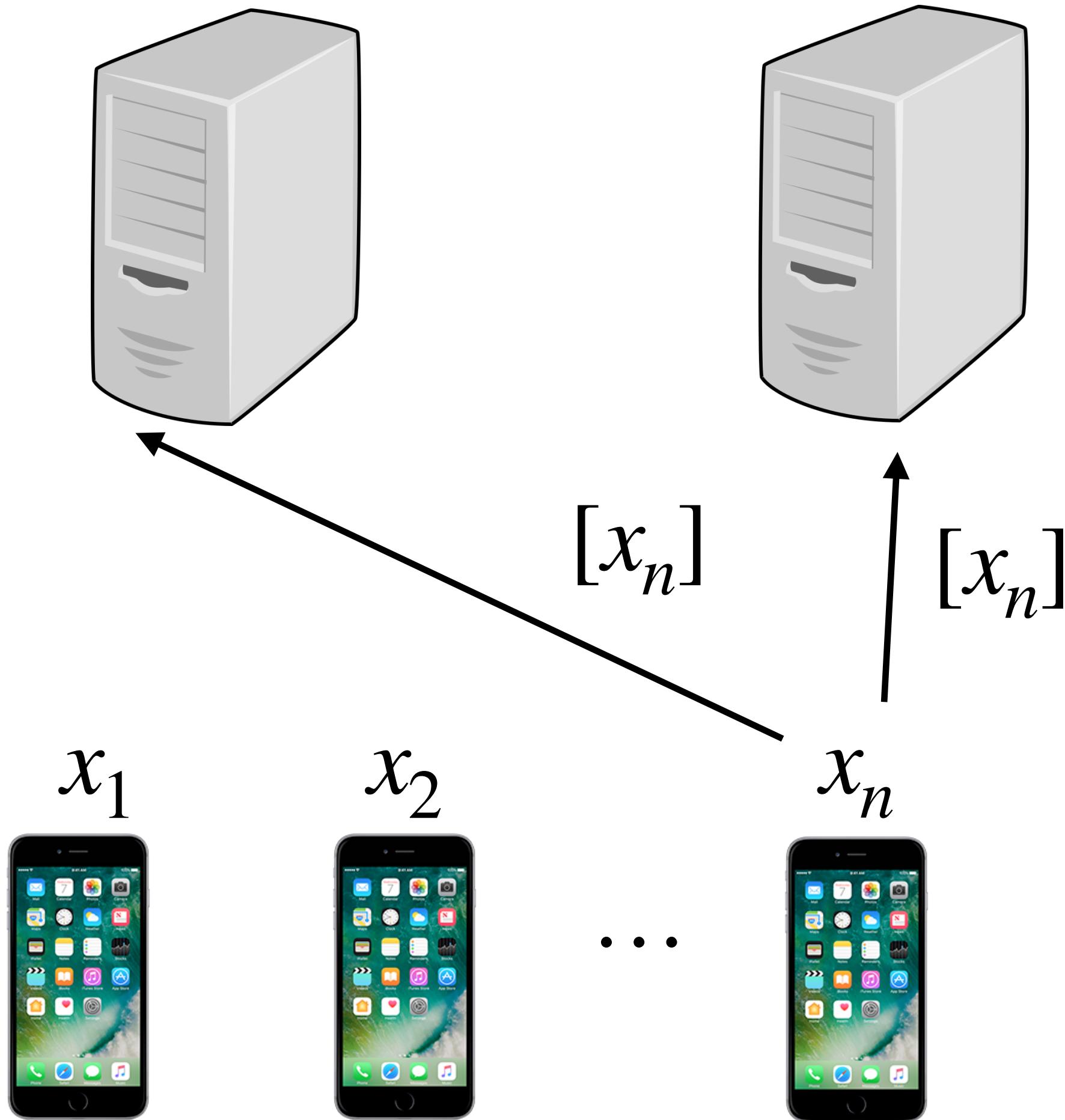
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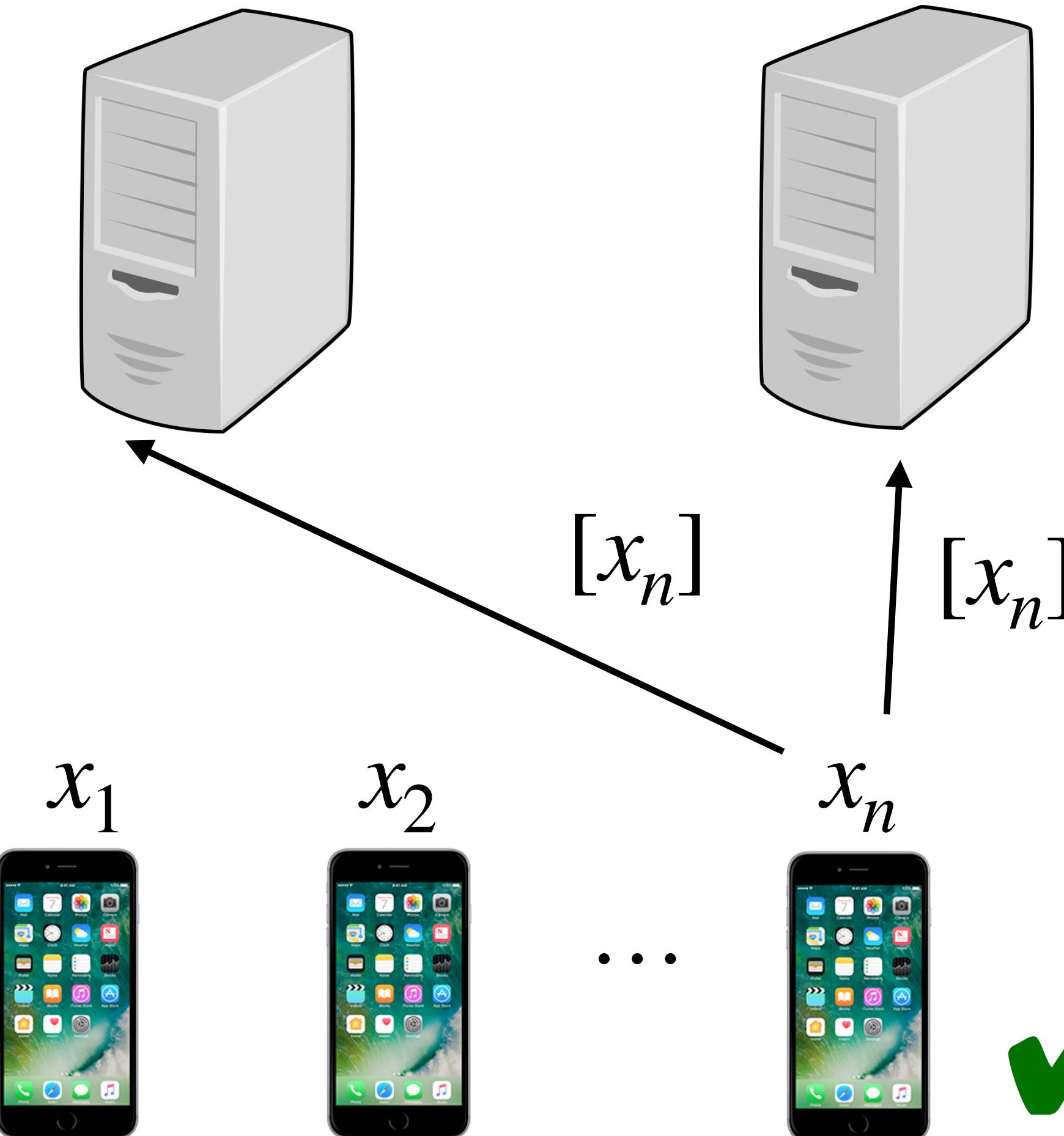
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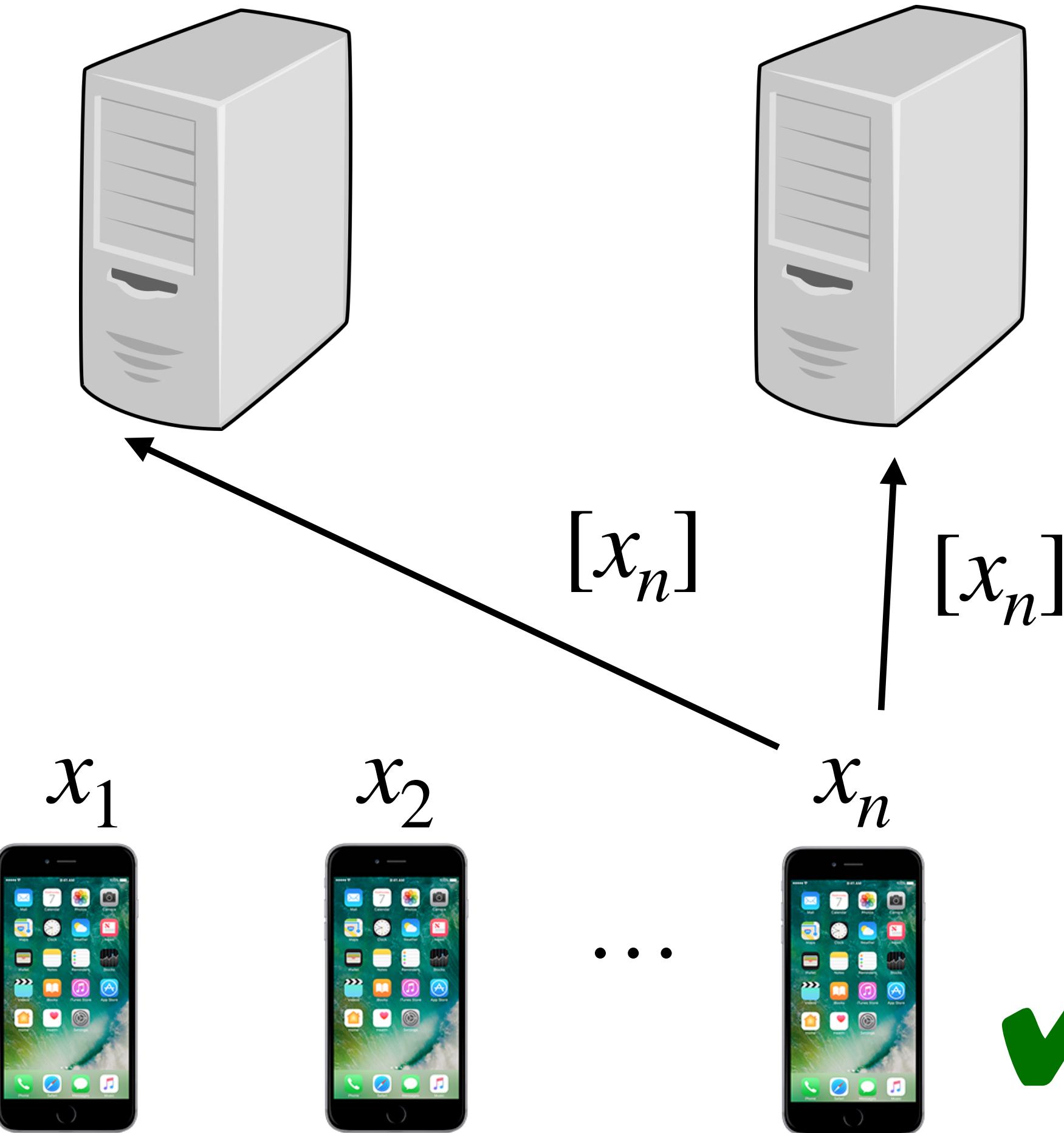
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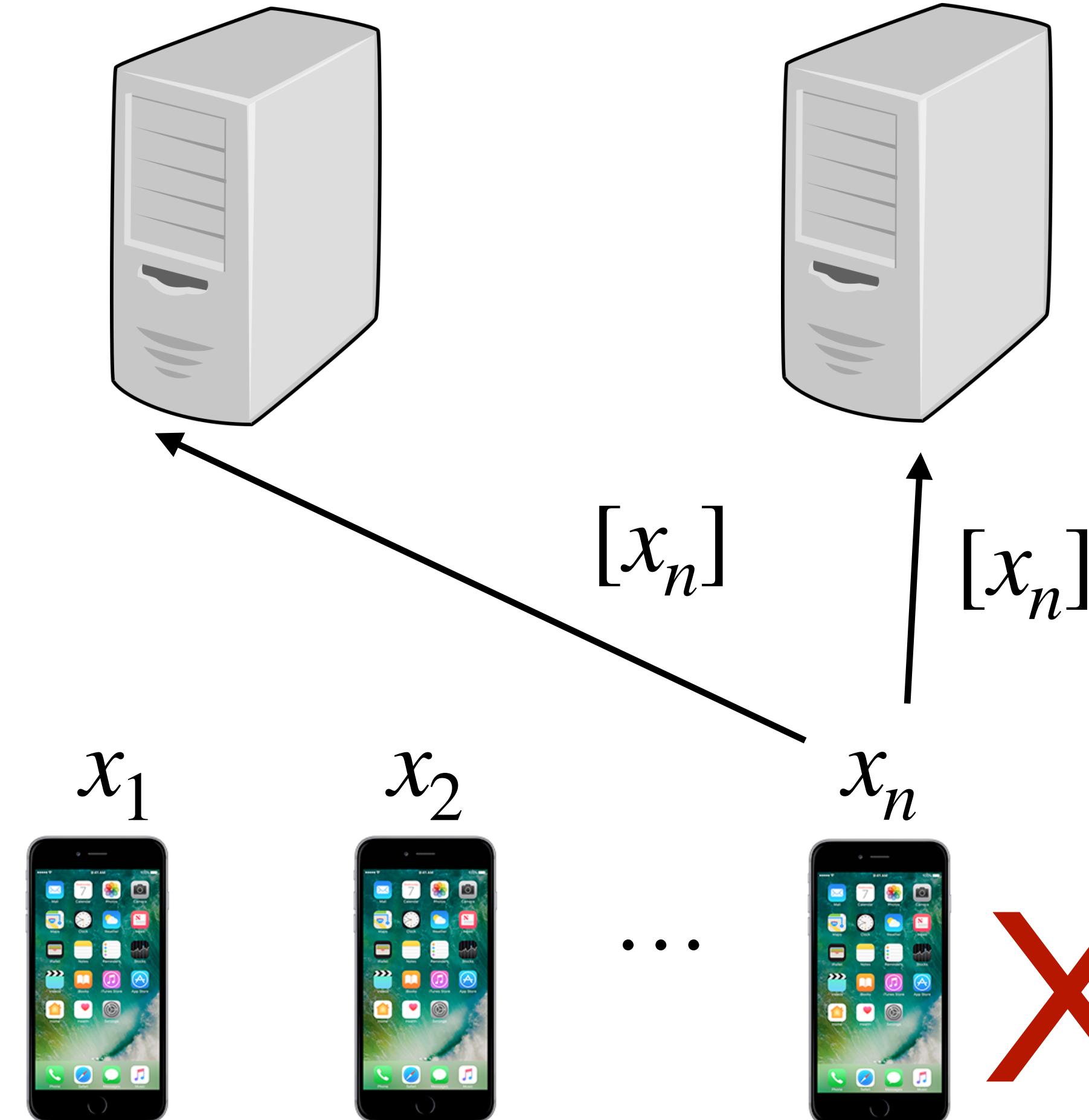
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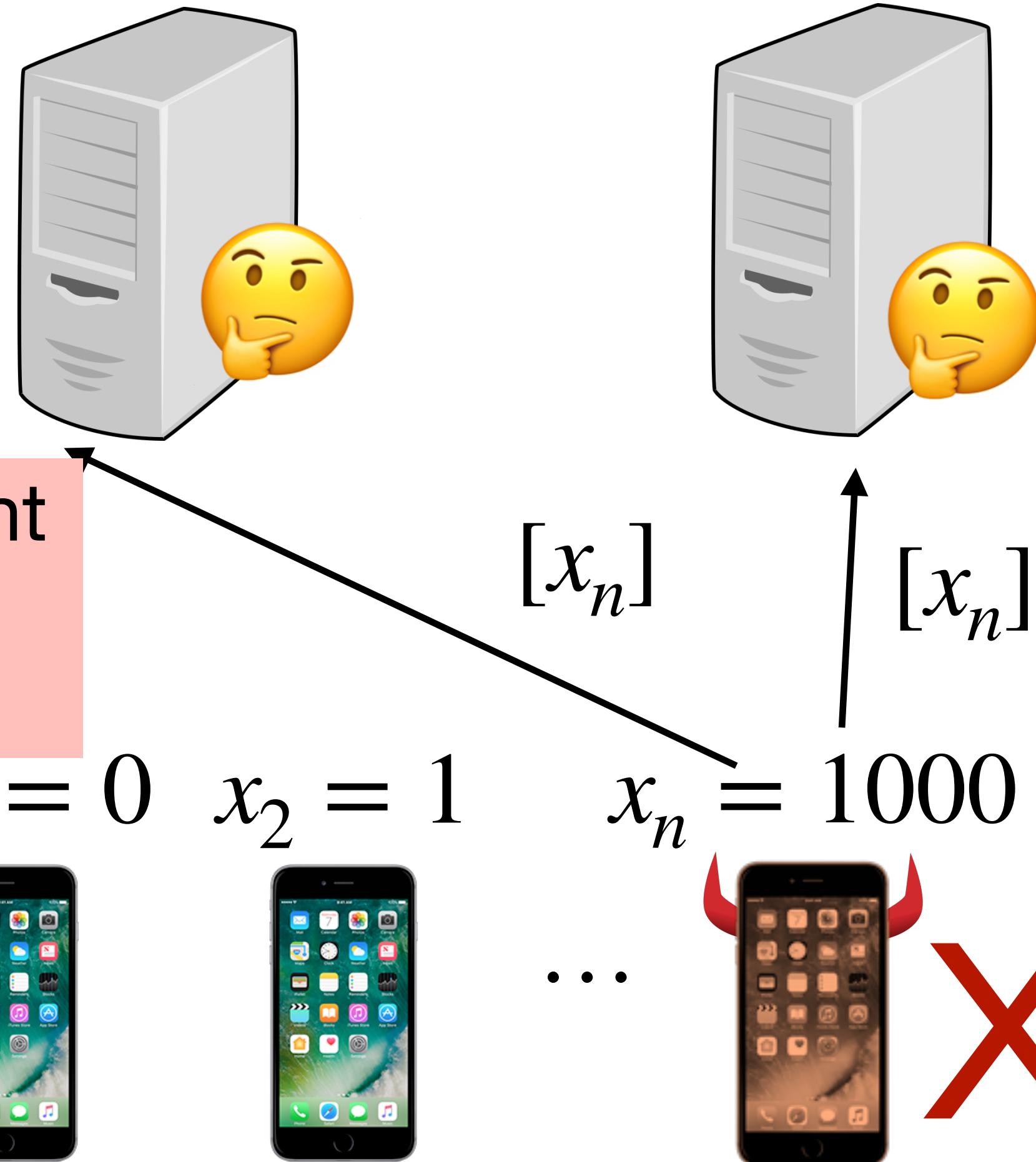
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Robustness: Malicious clients have bounded influence

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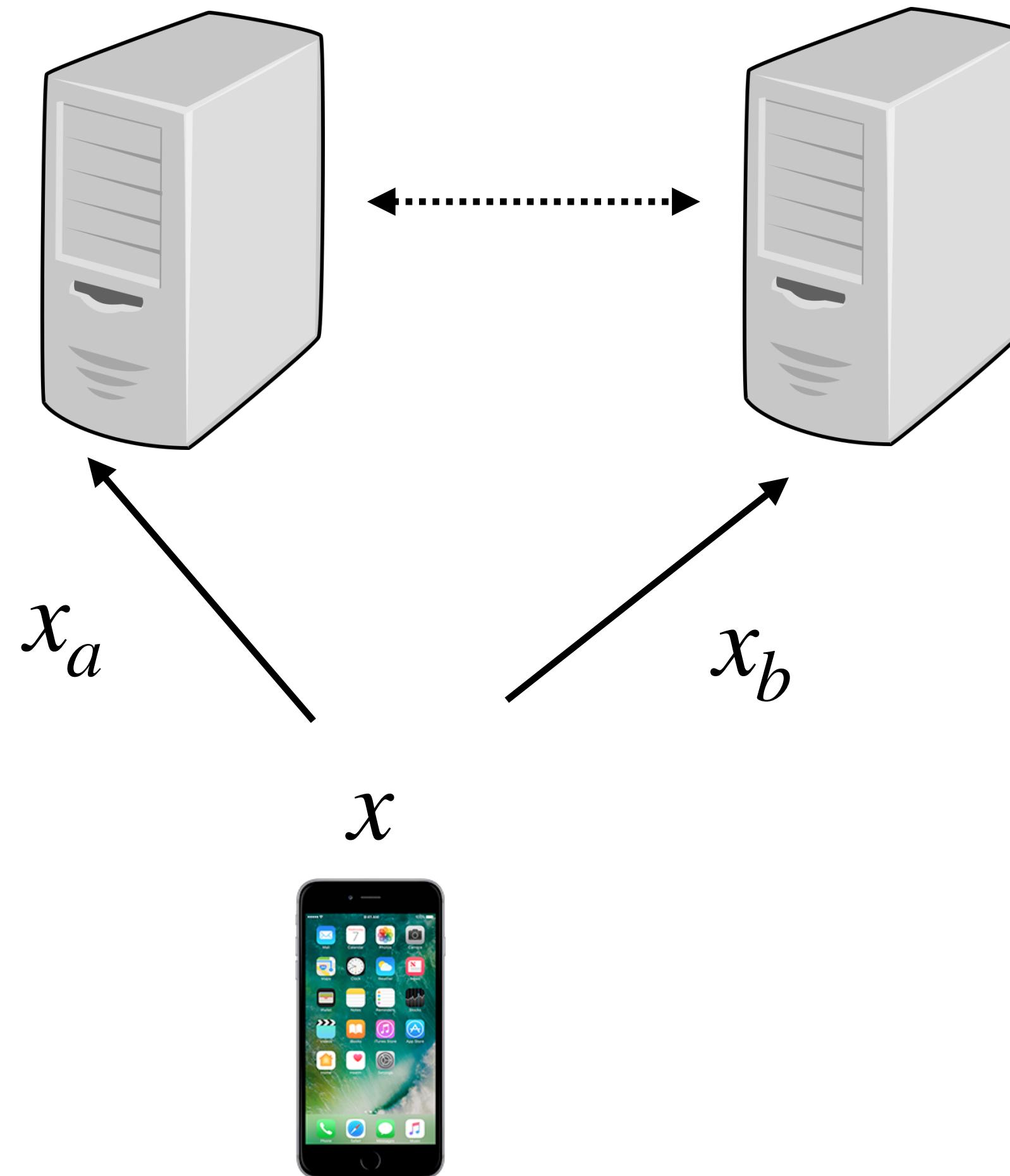
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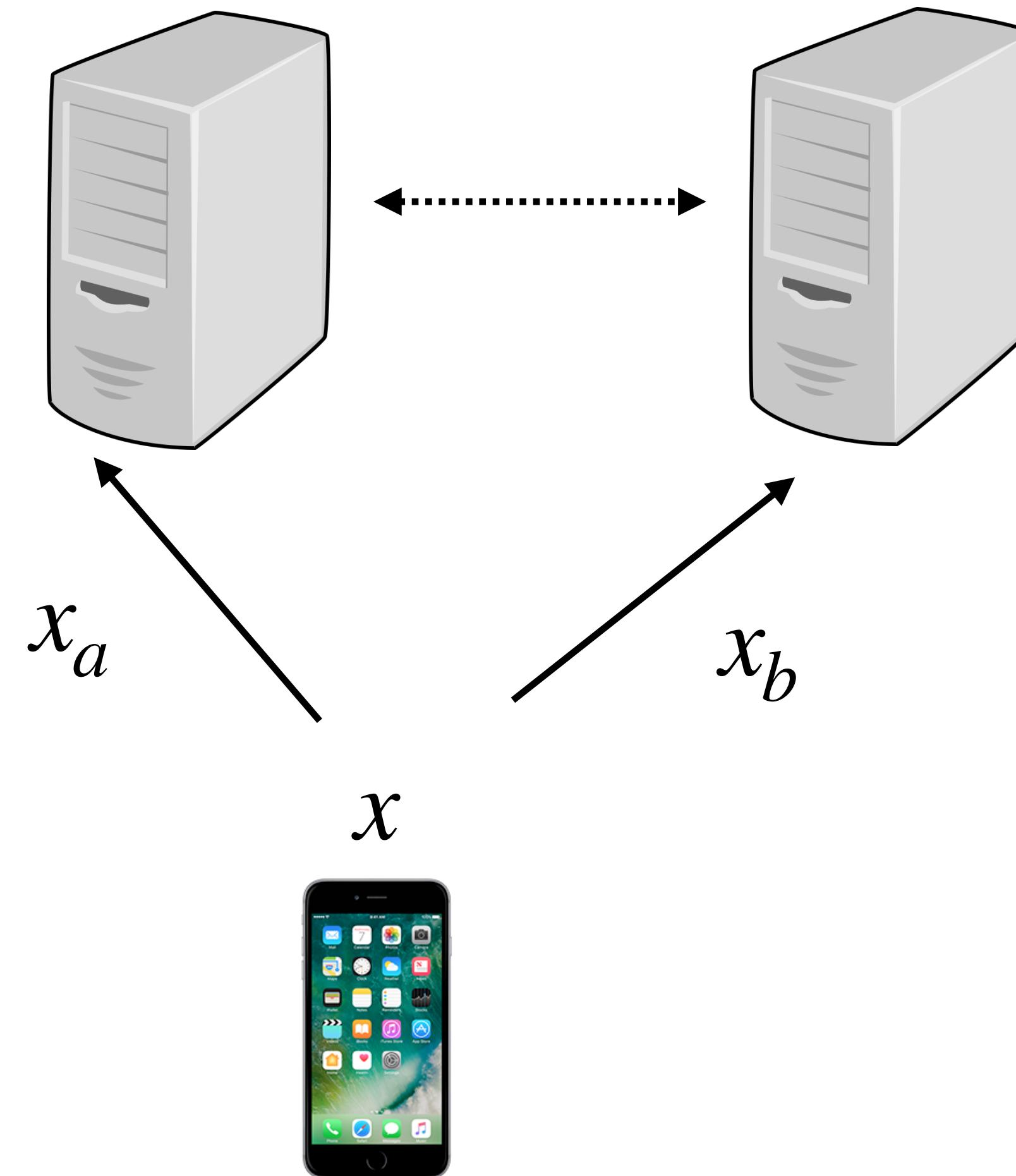
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Secret-shared non-interactive proofs (SNIPs)



Goal: Servers want to learn if $x_a + x_b \in \{0,1\}$ without learning x

Secret-shared non-interactive proofs (SNIPs)



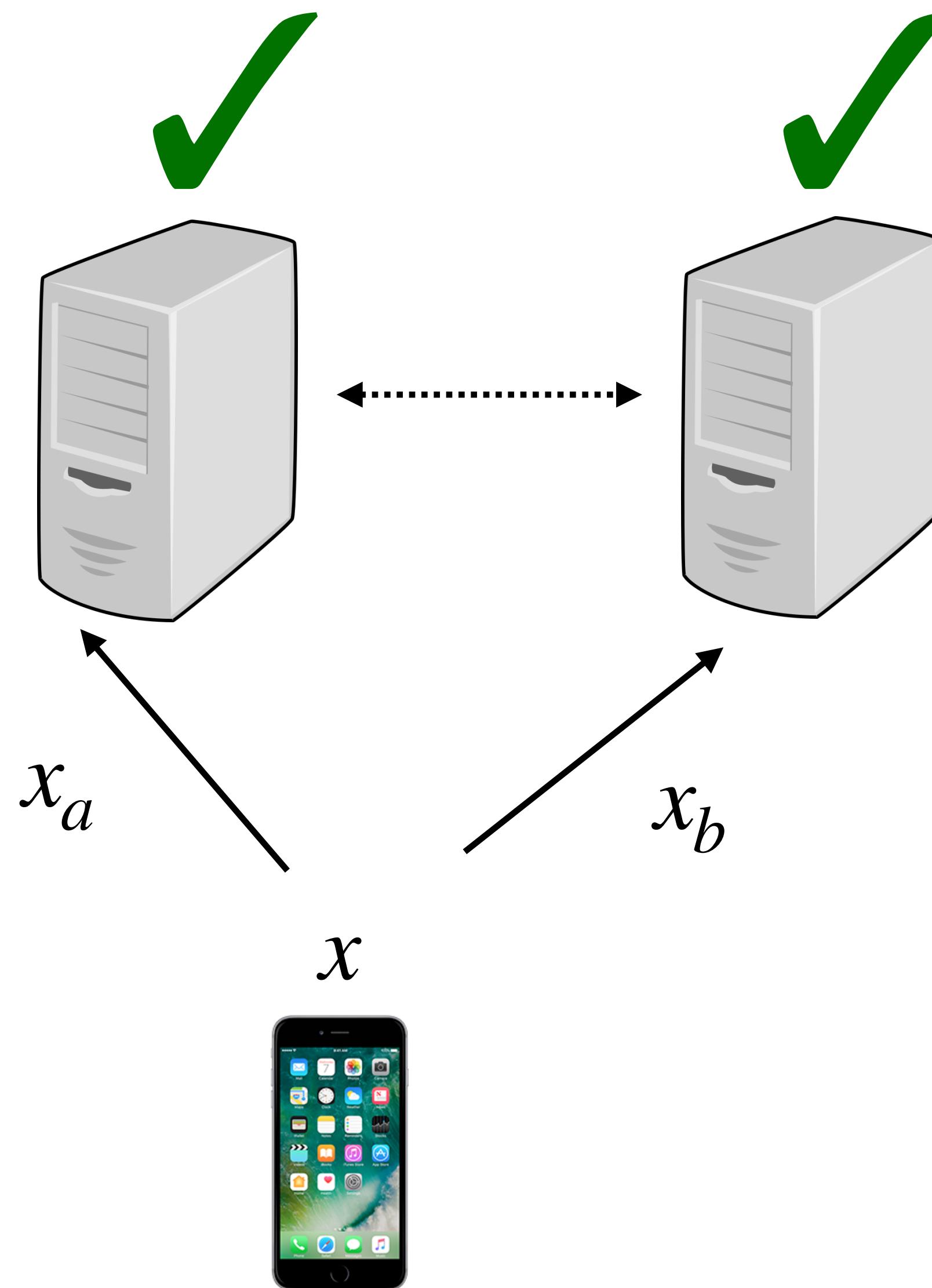
More generally, servers

- Hold shares of the client's private value x
- Hold an arbitrary public predicate $\text{Valid}(\cdot)$
- Want to test if $\text{Valid}(x) = 1$

EX: $\text{Valid}(x) = "x \in \{0,1\}"$

What other Valid predicates might be useful?

Secret-shared non-interactive proofs (SNIPs)

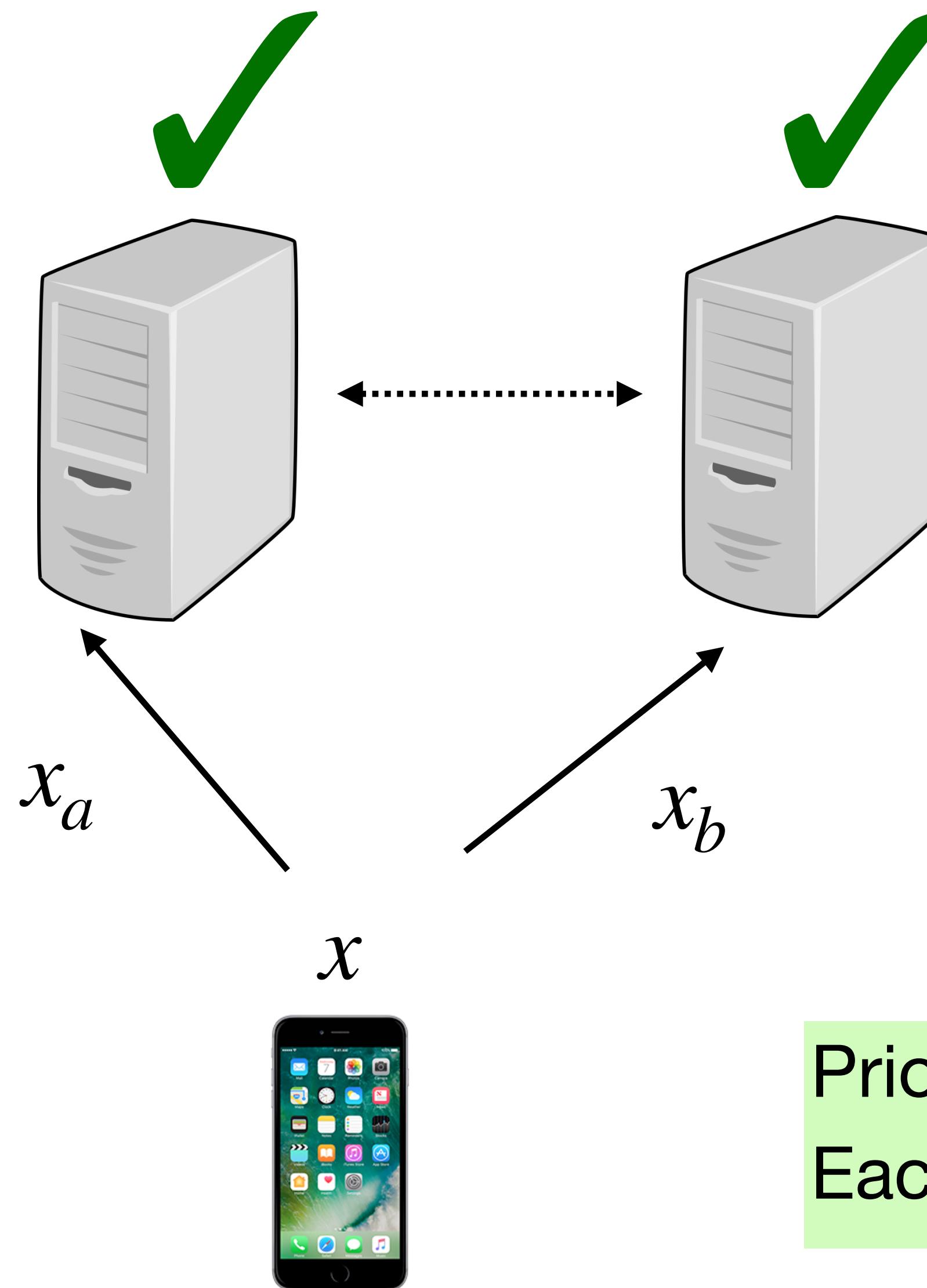


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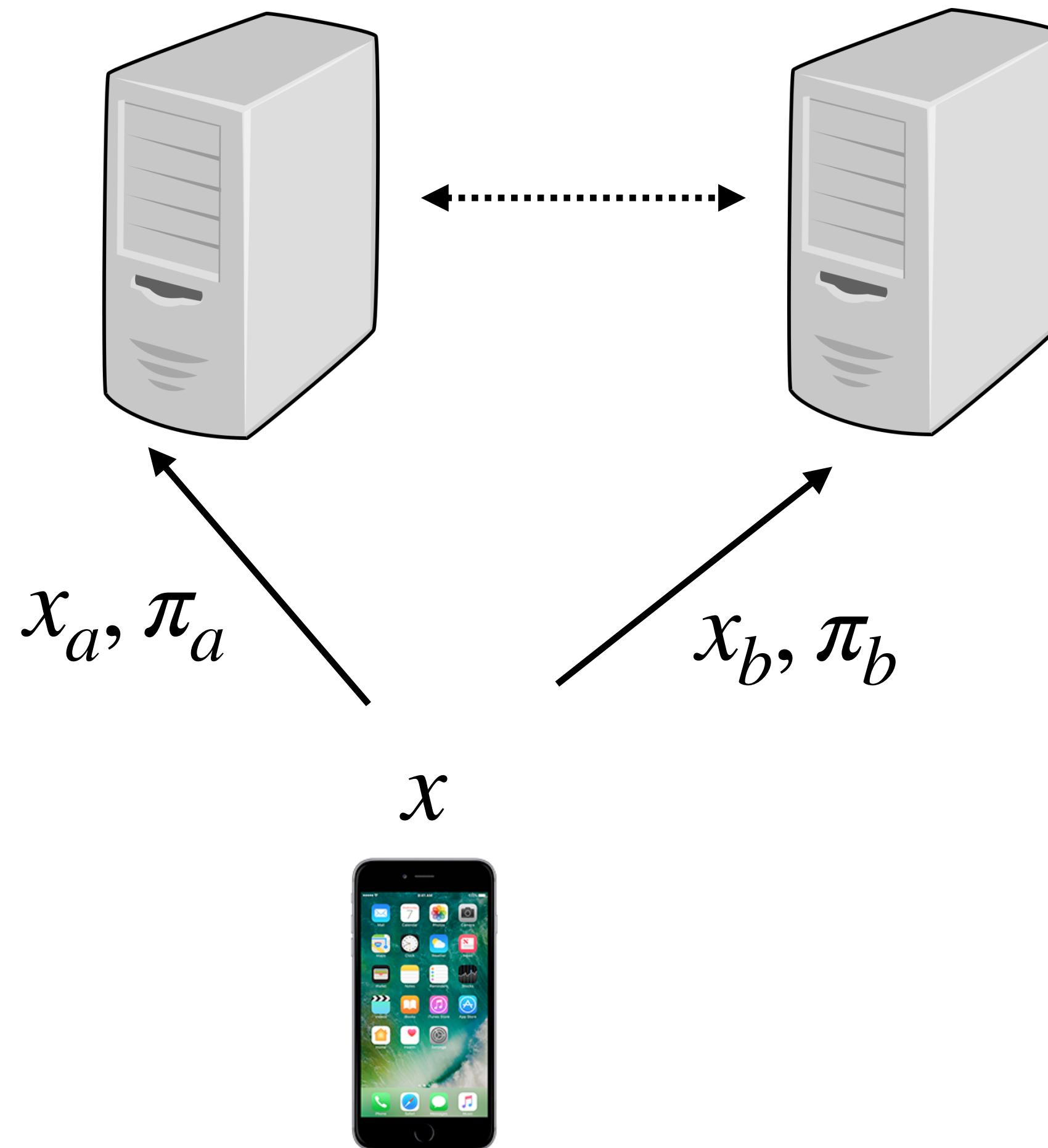
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Prio servers detect and reject malformed client input.
Each client can influence final sum by at most $+\text{-} 1$.

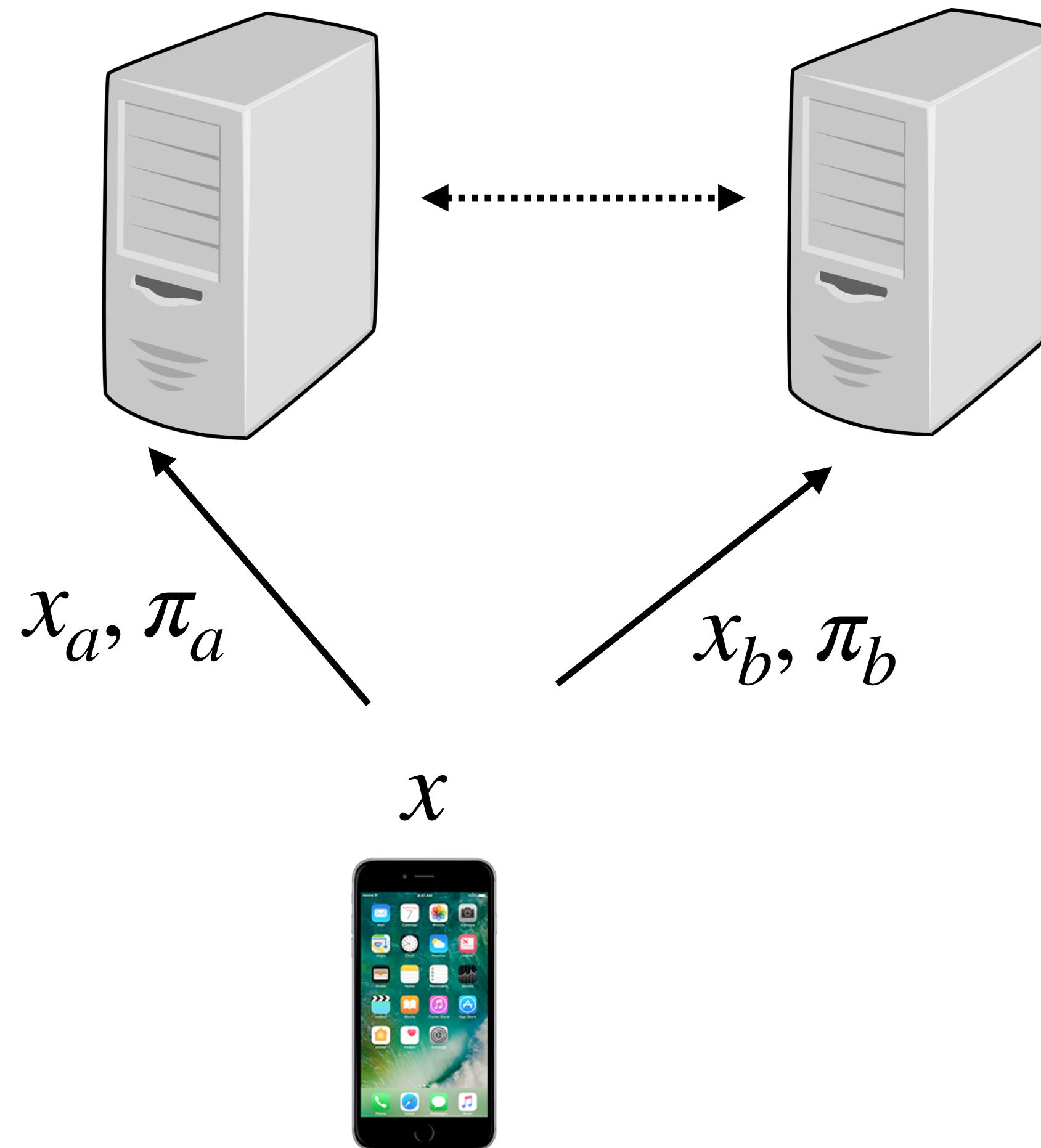
How do SNIPs work?



Servers could run MPC to check that
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Idea: Client can make servers' job easier!

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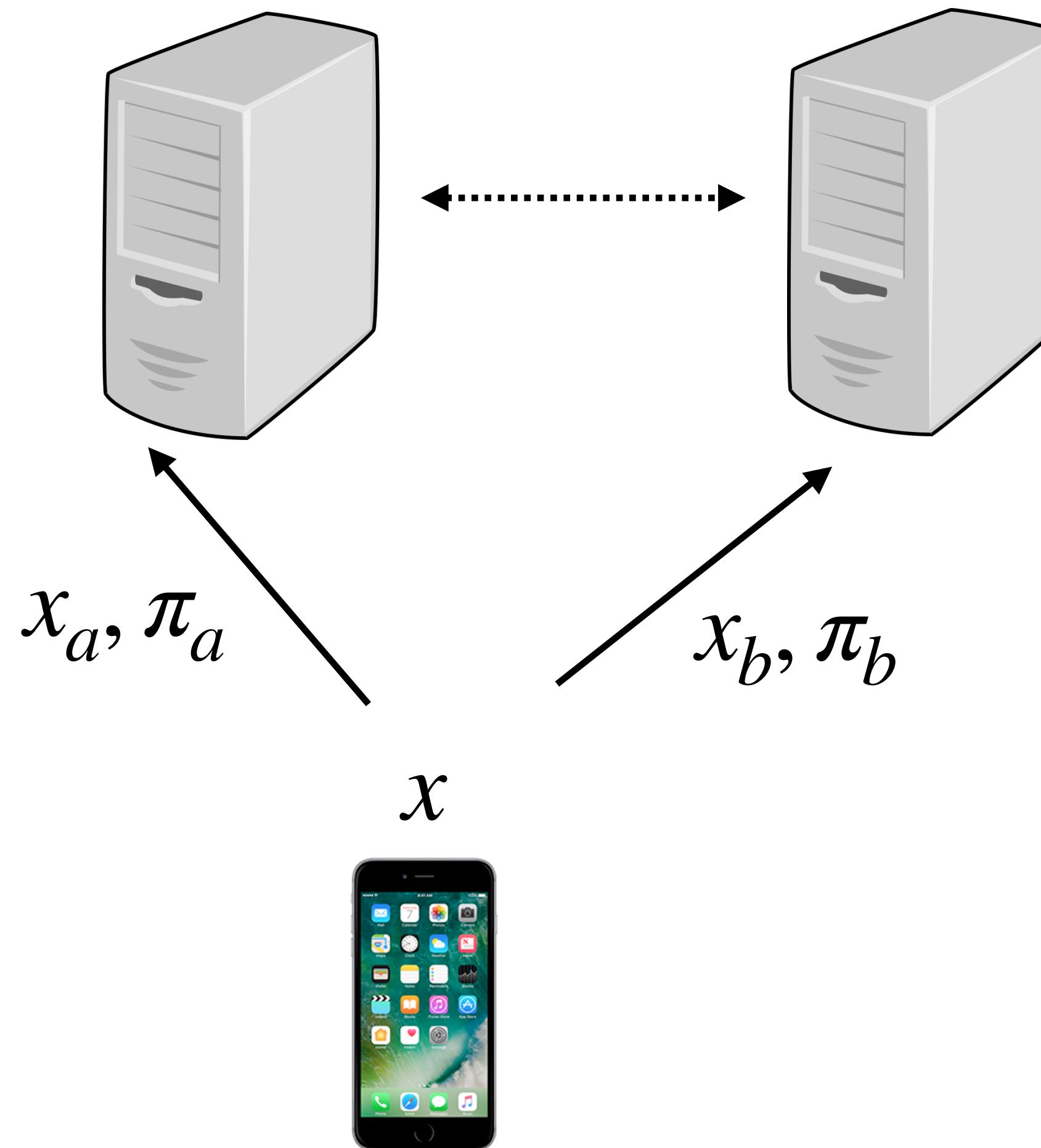


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- Client generates the transcripts that servers *would* have observed in MPC

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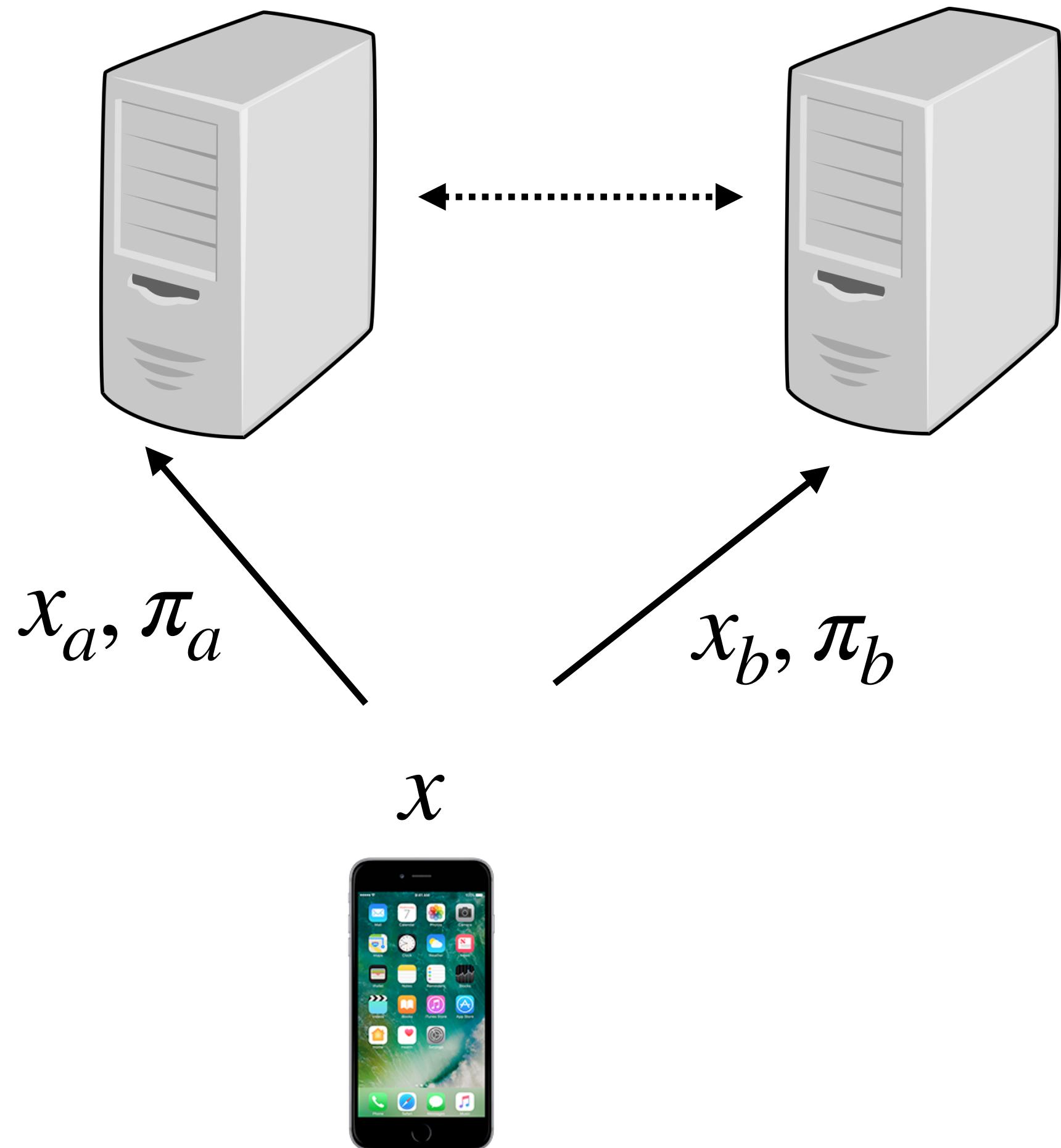
Servers check that the transcripts are valid and consistent.

- Checking a transcript is much easier than generating it

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Recall in MPC:

- Addition is free
- Multiplication requires communication

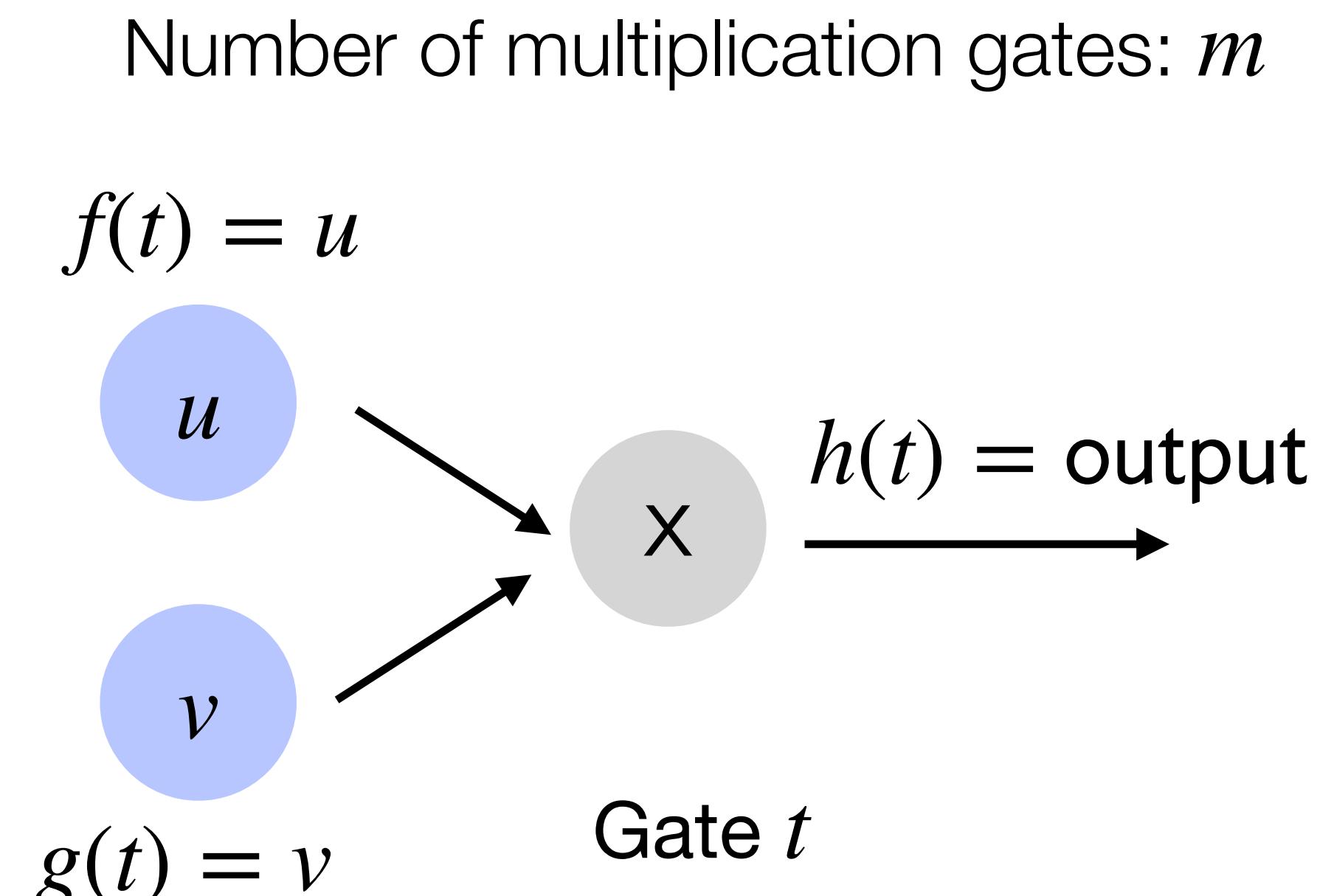
Goal: Use client to efficiently check multiplications

How do SNIPs work?

Client proof generation

1. Define 3 polynomials: f, g, h
2. For multiplication gate t with inputs u_t, v_t :
 - $f(t) = u_t$ ← Degree $m - 1$
 - $g(t) = v_t$ ← Degree $m - 1$
3. Let $h = f \cdot g$ ← Degree $2m - 2$
4. Secret-share coefficients of h to 2 servers

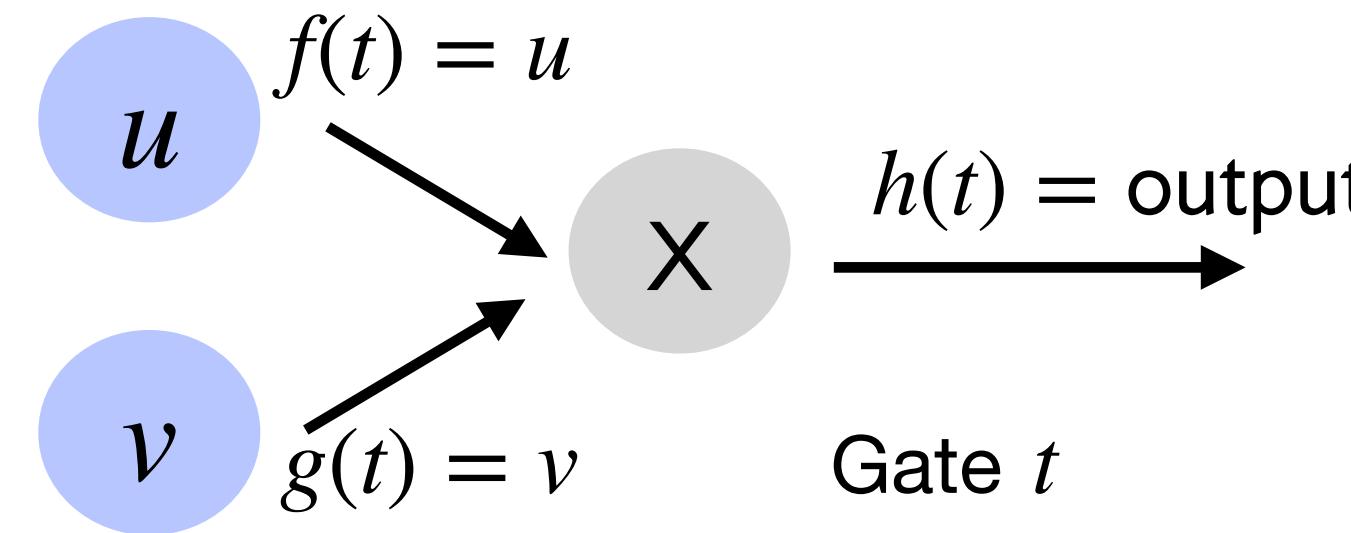
Goal: prove $\text{Valid}(x) = 1$



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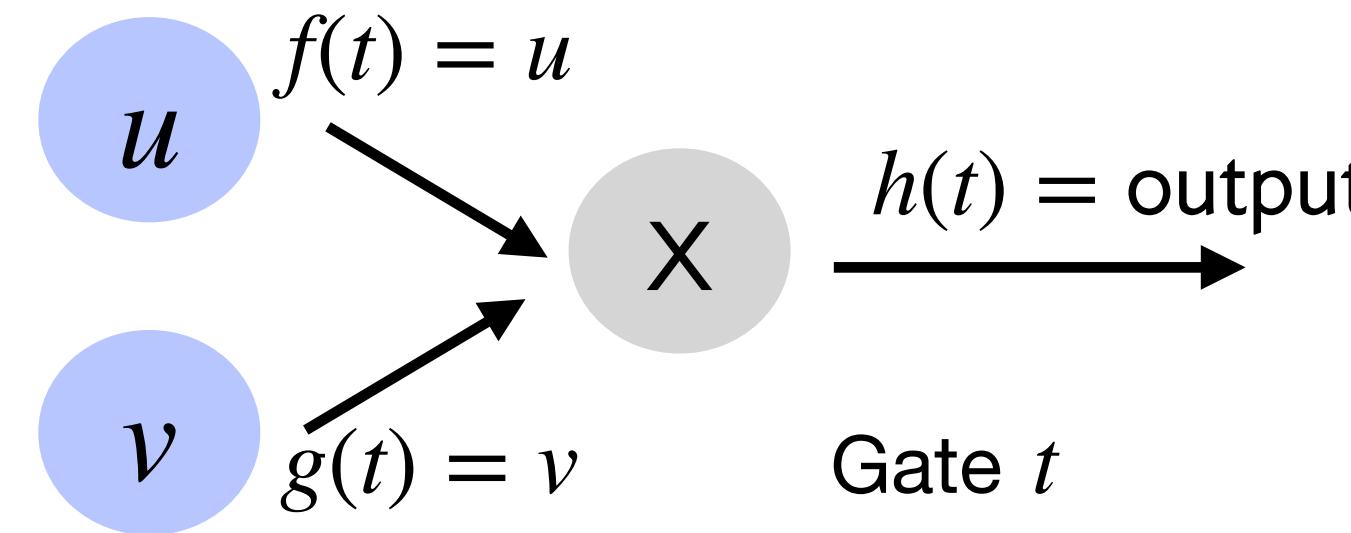
Server verification

1. Server can reconstruct secret shares for every wire value
 - Has share of input
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2. Servers can construct shares of polynomials f, g

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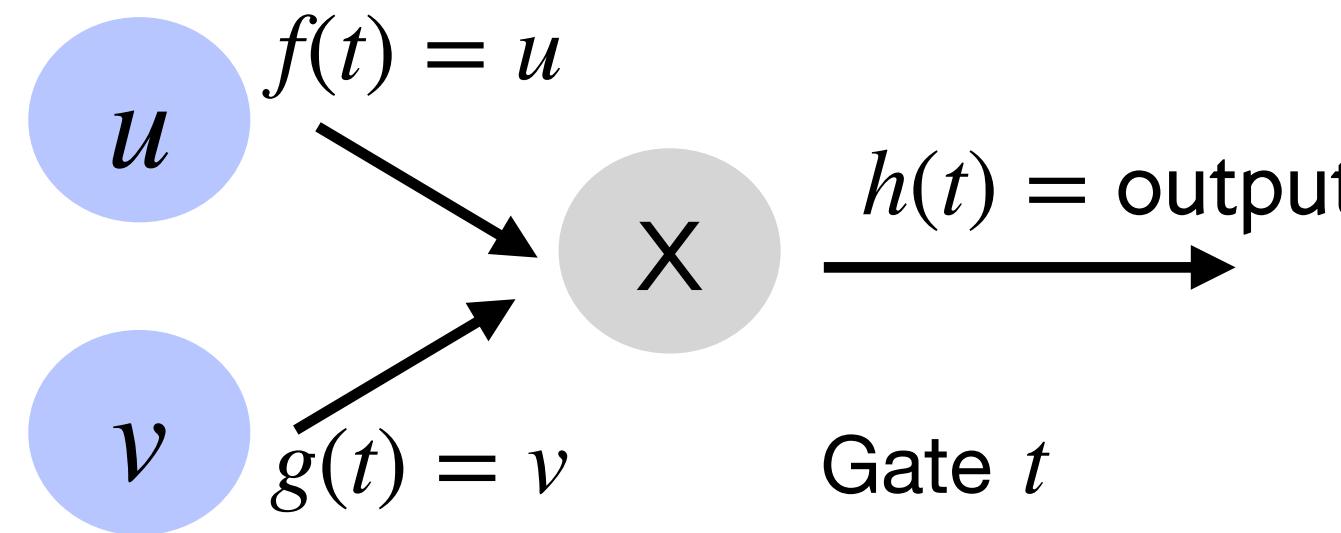
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Question: How can the servers tell if the client sent the wrong h ?

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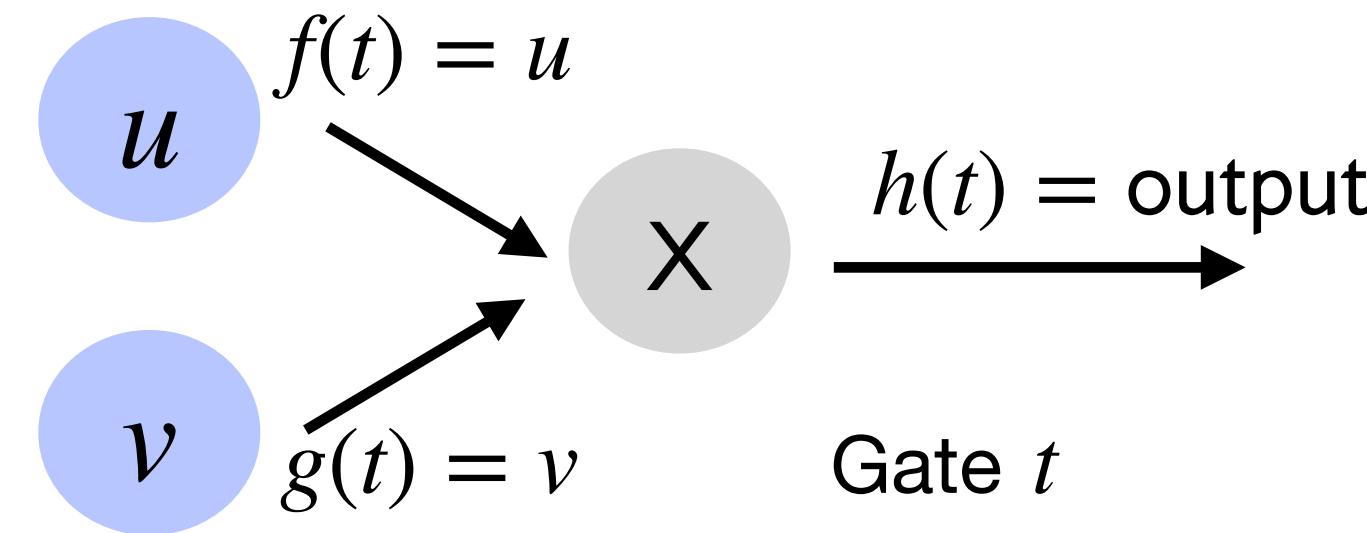
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3. Check that h is well-formed ($f \cdot g = h$)
 - Servers choose random r
 - Evaluate $\alpha \leftarrow f(r) \cdot g(r) - h(r)$
 - If $f \cdot g \neq h$, $\Pr[\alpha = 0]$ is negligible
 - Checking this requires 1 multiplication of secret-shared values

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Probability that client cheated and didn't get caught is negligible

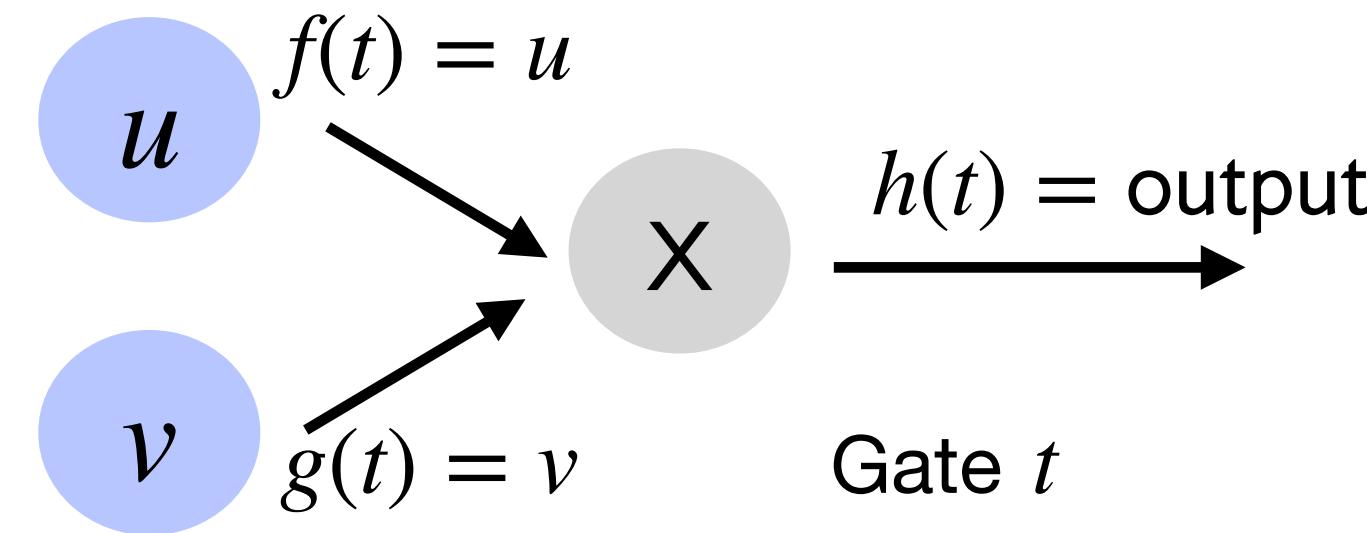
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Due to the fact that polynomial of at most degree $2m - 2$ can have at most $2m - 2$ zeroes (m is number of multiplication gates)

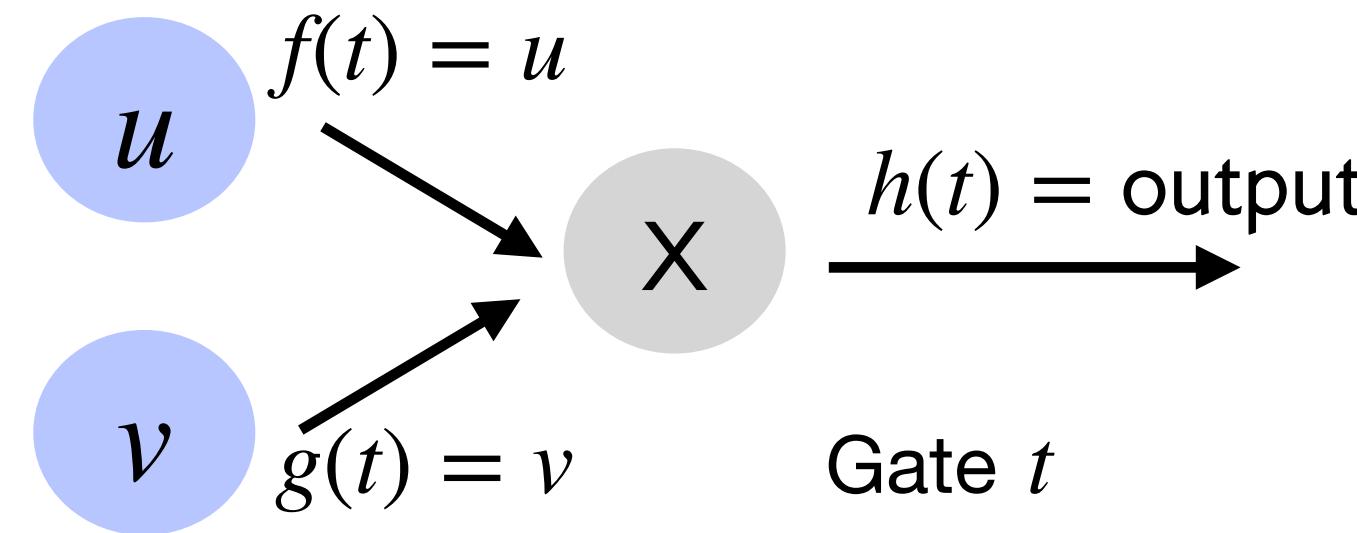
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Client can provide Beaver triple to servers.

- If client sends malformed Beaver triple, servers still catch cheating client with overwhelming probability

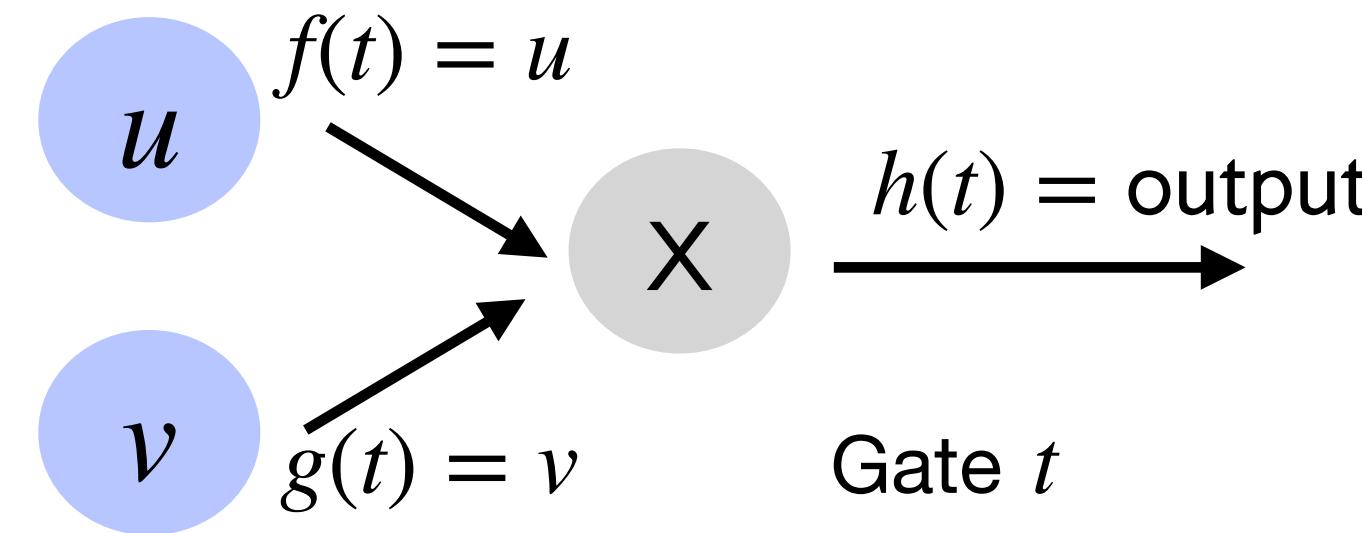
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 - Checking this requires 1 multiplication of secret-shared values
4. Check output wire is shares of 1 to verify $\text{Valid}(x) = 1$

What can Prio compute with sums?

If you can compute private sums, you can compute many other interesting aggregates using known techniques

- Average
- Variance
- Standard deviation
- Most popular (approx)
- Frequency counts for a small set
- Min and max (approx)
- Least-squares regression
- Stochastic gradient descent [Bonawitz et al. 2016]

Outline

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References

Bonawitz, Keith, Vladimir Ivanov, Ben Kreuter, Antonio Marcedone, H. Brendan McMahan, Sarvar Patel, Daniel Ramage, Aaron Segal, and Karn Seth. "Practical secure aggregation for federated learning on user-held data." *arXiv preprint arXiv:1611.04482* (2016).

Corrigan-Gibbs, Henry, and Dan Boneh. "Prio: Private, robust, and scalable computation of aggregate statistics." In *14th USENIX symposium on networked systems design and implementation (NSDI 17)*, pp. 259-282. 2017.

Prio slides from Henry Corrigan-Gibbs: <https://people.csail.mit.edu/henrycg/files/academic/pres/nsdi17prio-slides.pdf>