



# COM-506

## Prio: Private, Robust, and Scalable Computation of Aggregate Statistics

Federico Villa, Gabriele Stentella, Octave Charrin

# Context

- Many modern devices collect data and send it to cloud services.
- Storing private data, the services create a single point of failure.
- Huge threat for privacy and security.
- The services need aggregate statistics.

Collect data from  
mobile apps.

Private compute  
services.

Spread data over  
multiple countries.

How do we split trust in a way that protects  
privacy **and** maintains functionality?

# Introduction

**Idea:** the clients send an encrypted share of their data points to each aggregator.

*How?*

## Goals:

1. Servers learn the output of the aggregation function (correctness).
2. But learn nothing more (privacy).
3. The system is robust  $\Rightarrow$  detects incorrect submissions.
4. The protocol is efficient and scalable  
 $\Rightarrow$  no heavy public-key cryptography operations.

# Previous approaches

## Randomized response

- Clients flip their bits with fixed probability  $p < 0.5$
- Every bit leaks information (especially for low  $p$ ).  $\Rightarrow$  weak privacy
- With  $p$  too high the aggregation becomes useless.
- Bounded client contribution.

## Encryption

- Stronger privacy guarantees.
- Unbounded client contribution.
- Not scalable.

# Prio - overview

- Small number of servers, large number of clients.
- Built using Secret-shared Non-Interactive Proofs (SNIPs) and Affine-aggregatable Encodings (AFEs).

## Assumptions on the network

- PKI and basic cryptographic primitives.
- No synchrony.
- Adversary monitors the network and controls the packets.

# Prio - simplified

We use  $[x]_s$  to denote the sth share of  $x$ :

$$x = \sum_s [x]_s$$

**Input:** one bit integer  $x_i$

**Aggregation:** sum  $\sum_i x_i$

1. Private value secret-shared between  $s$  servers.  $x_i = [x_i]_1 + \dots + [x_i]_s \in \mathbb{F}_p$
2. Each server add the share to its internal accumulator.
3. The servers publish the accumulators.
4. The sum of the accumulators is the desired aggregation.

- Privacy from secret sharing.
- **No robustness.**
- Only sum.

# SNIPs: Secret-shared Non-Interactive Proofs

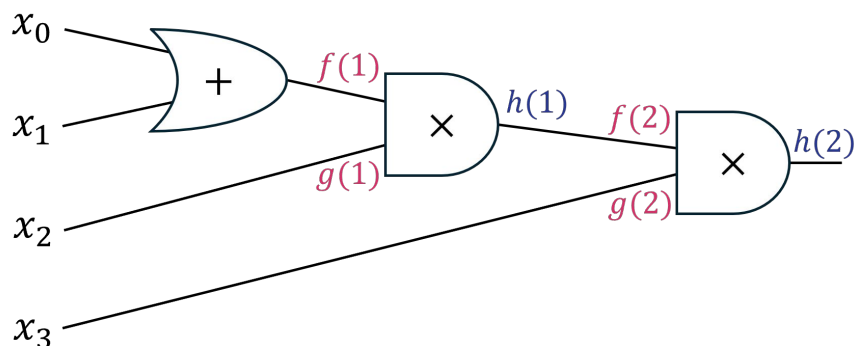
- Linear additive secret-sharing over field  $\mathbb{F}$
- Validation predicate **Valid**  $\Rightarrow$  encoded in an arithmetic circuit

## SNIP protocol

1. Client evaluates the circuit.
2. Servers check consistency.
3. Polynomial validation  $\Rightarrow$  polynomial identity test.  
Multiplication of shares.
4. Final computation and verification.

# 1. Client evaluates the circuit

- Three randomized polynomials  $f, g, h$ .
- $M$  multiplication gates.
- Left input  $u_t$
- Right input  $v_t$
- $h(t) = f(t) \cdot g(t) = u_t \cdot v_t \quad \forall t \in \{1, \dots, M\}$
- $u_0, v_0 \sim \mathbb{F}$



Output

$[f(0)]_i$     $[g(0)]_i$     $[h]_i$

shares of the coefficients of  $h$



## 2. Servers check consistency

- Internal derivation of values  $[f]_i, [g]_i$
- If all parties are honest:  $f \cdot g = h$
- In case of malicious client:  $\hat{h} \neq \hat{f} \cdot \hat{g}$

## 3. Polynomial validation

**Goal:** Detect with high probability a cheating client.

1. Sample a random value from the field.
2. Evaluate polynomials on the random value.
3. Get shares of  $\sigma = r \cdot (\hat{f}(r) \cdot \hat{g}(r) - \hat{h}(r))$
4. Check the sum of those shares is 0.

If  $\hat{h} \neq \hat{f} \cdot \hat{g}$  then the polynomial represented by  $\sigma$  is of degree at most  $2M + 1$ : with random evaluation, we detect the cheat with probability

$$\geq 1 - \frac{2M+1}{|\mathbb{F}|}$$

### Beaver's Multi-Party Computation

Clients choose the triple  $(a, b, c) \in \mathbb{F}^3$  and send shares to the servers.

## 4. Final computation and verification

- Share the values of the shares of the output of **Valid**
- Check that they sum up to 1.

$$\textbf{SNIP proof tuple} \quad \pi = (f(0), g(0), h, \underbrace{a, b, c}_{\text{Beaver's triple}})$$

### Efficiency

- Server-to-server communication cost same as local cost of circuit evaluation.
- Client-to-server communication linear in the size of the circuit.

# Desired Properties of a useful SNIP

- **Correctness:** If all parties are honest, the servers will accept  $x$ .
- **Soundness:** If all servers are honest, and if  $\text{Valid}(x) \neq 1$ , then the servers will almost always reject  $x$ , no matter how the client cheats.

# Formal definition

1. Run the adversary  $\mathcal{A}$ . For each server  $i$ , the adversary outputs a set of values:

- $[x]_i \in \mathbb{F}^L$ ,
- $([f(0)]_i, [g(0)]_i) \in \mathbb{F}^2$ ,
- $[h]_i \in \mathbb{F}_{2M}[X]$  of degree at most  $2M$ , and
- $([a]_i, [b]_i, [c]_i) \in \mathbb{F}^3$ .

2. The *Master server* chooses a random  $r \xleftarrow{\$} \mathbb{F}$ . Each server compute their shares  $[f]_i$  and  $[g]_i$  as in the real protocol, and evaluate  $[f(r)]_i$ ,  $[r \cdot g(r)]_i$ ,  $[r \cdot h(r)]_i$ , and  $[h(M)]_i$ .

3. The servers compute  $h(M) = \sum_i [h(M)]_i$ , and

$$\sigma = r \cdot (f(r)g(r) - h(r)) + (c - ab)$$

4. We say that the adversary wins the game if:

$$h(M) = 1, \quad \sigma = 0, \quad \text{and} \quad \text{Valid}(x) \neq 1$$

Soundness:

$$\Pr[\mathcal{A} \text{ Wins}] \leq \frac{2M + 1}{|\mathbb{F}|}$$

$$\sigma = P(r)$$

A Wins if:

$$h(M) = 1, \quad \sigma = 0, \quad \text{and} \quad \text{Valid}(x) \neq 1$$

$$P(t) = t \cdot Q(t) + (c - ab)$$

$$Q(t) = f(t)g(t) - h(t)$$

**Case  $fg \neq h$  :**

- $P$  is a non-zero polynomial of degree at most  $2M+1$ .
- The choice of  $r$  is independent of  $(a, b, c)$  and  $Q$ , since the adversary must produce these values before  $r$  is chosen.

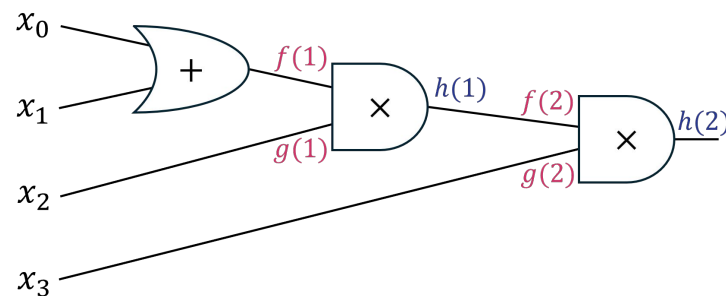
$\Rightarrow$  The choice of  $r$  is independent of  $P$ .

- $P$  has at most  $2M+1$  zeros in  $\mathbf{F}$ .

$$\Rightarrow \Pr[ P(r) = \sigma = 0 ] \leq (2M+1)/|\mathbf{F}|$$

**Case  $fg = h$  :**

- By induction:  $h(M) = \text{Valid}(x)$  (wlog assume that the circuit ends with a multiplication gate)



$$\Rightarrow \Pr[ h(M) = 1 \text{ and } \text{Valid}(x) \neq 1 ] = 0$$

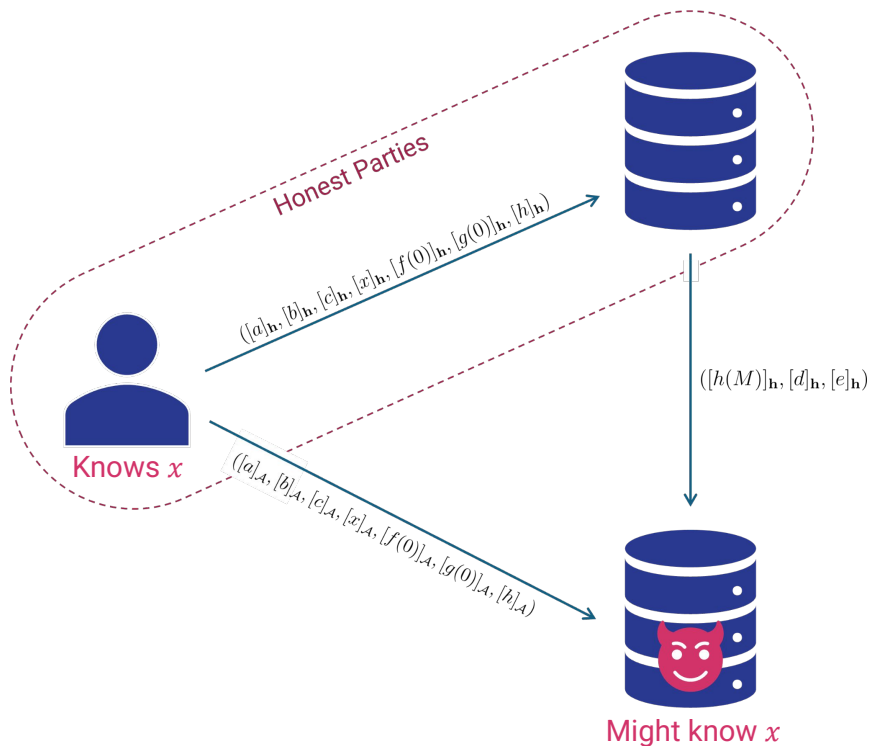
In both cases:  $\Pr[\mathcal{A} \text{ Wins}] \leq \frac{2M+1}{|\mathbf{F}|}$

# Desired Properties of a useful SNIP

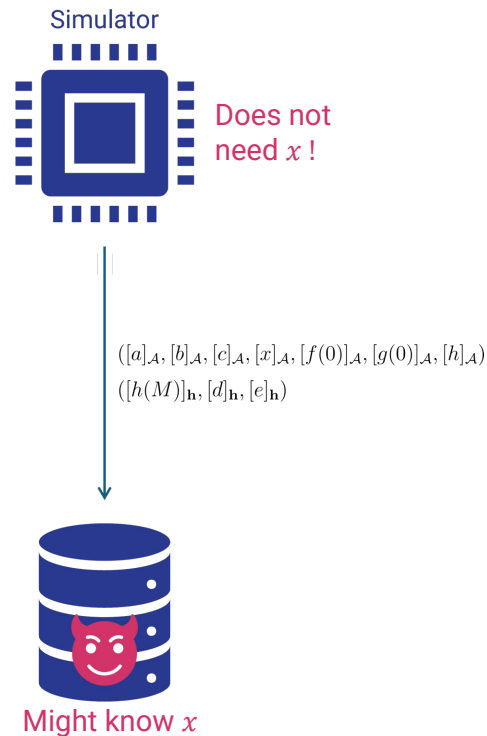
- **Correctness:** If all parties are honest, the servers will accept  $x$ .
- **Soundness:** If all servers are honest, and if  $\text{Valid}(x) \neq 1$ , then the servers will almost always reject  $x$ , no matter how the client cheats.
- **Zero knowledge:** If the client and at least one server are honest, then the servers learn nothing about  $x$ , except that  $\text{Valid}(x) = 1$ .

# Zero Knowledge - Proof Sketch

“Real World”



“Ideal World”



# Zero Knowledge - Proof Sketch

In this game, the adversary tries to distinguish the two worlds.

- The simulator generates the initial adversary view **at random**.
- We can show that **the two views are distributed identically**.

(random sampling of  $r$ ,  $f(0)$  and  $g(0)$  in the real world + hiding from secret sharing)

- Since the simulator does not know  $x$ :

⇒ Participating in the SNIP gives no extra information about  $x$ .



# Affine-aggregatable encodings (AFEs)

So far we can:

- Compute private sums over client-provided data (Secret-sharing)
- Check arbitrary validation predicate against data (SNIP)

*How can we compute more complex statistics ?*

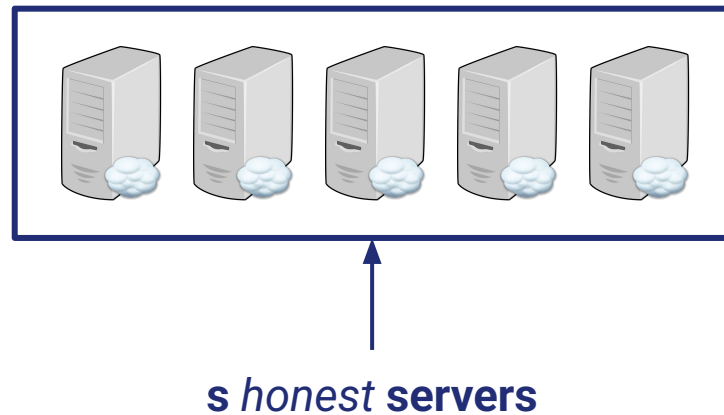
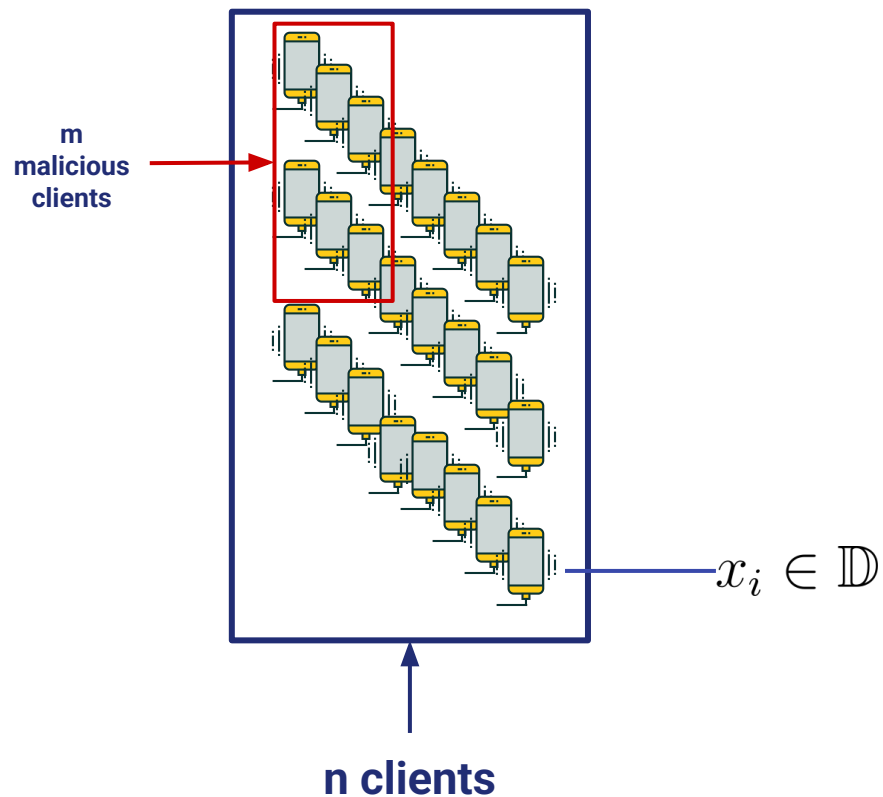
**Idea:** Encode private data to make the statistic computable over the sum of encoding.

# AFE concrete example

Computing the variance of  $b$ -bit integers:  $\text{Var}(X) = \mathbb{E}[X^2] - \mathbb{E}[X]^2$

- $\text{Encode}(x) = (x, x^2, \beta_0, \beta_1, \dots, \beta_{b-1})$  ← Secret-sharing
- $\text{Valid}(\text{Encode}(x)) = \left( x = \sum_{i=0}^{b-1} 2^i \beta_i \right) \wedge (x \cdot x = x^2) \wedge \bigwedge_{i=0}^{b-1} [\beta_i \cdot (\beta_i - 1) = 0]$  ← SNIP
- $(\sigma_0, \sigma_1) = \sum_{i=1}^n \text{Trunc}_2(\text{Encode}(x_i)) = \sum_{i=1}^n (x_i, x_i^2) = \left( \sum_{i=1}^n x_i, \sum_{i=1}^n x_i^2 \right)$
- $\text{Decode}(\sigma) = \frac{1}{n} (\sigma_1 - (\sigma_0)^2)$

## 7. Prio Protocol - Setting



$$f : \mathbb{D}^{n-m} \rightarrow \mathbb{A}$$

## 7. Prio Protocol - 4 steps

$$x_i \in \mathbb{D}$$

### 1. Upload phase

$$y_i \in \mathbb{F}^k$$

- input encoded using Affine-Aggregatable Encoding
- AFE encoded vector is split into secret shares
- SNIP proof is generated to prove data is well formed
- input shares and SNIP proof are sent to the servers

$$y_i \leftarrow \text{Encode}(x_i)$$

$$[y_i]_1, [y_i]_2, \dots, [y_i]_s$$

$$y_i = [y_i]_1 + [y_i]_2 \dots + [y_i]_s$$

# 7. Prio Protocol - 4 steps

## 2. Validation phase

- servers jointly verify the SNIP proofs received
  - rejects not well-formed submission
  - does not reveal information about the underlying data (except validity)
  - ensures robustness against malformed/malicious submissions

$$x_i \in \mathbb{D} \iff \text{Valid}(x_i) = 1$$

## 7. Prio Protocol - 4 steps

### 3. Aggregation phase

- each server initializes an accumulator to zero:
- for every valid client submission increments the accumulator
  - only truncated version of the client share carry necessary information

$$A_j \in F^{k'}$$

$$A_j \leftarrow 0$$

$$A_j \leftarrow A_j + \text{Trunc}_{k'}([y_i]_j) \in \mathbb{F}^{k'}$$

## 7. Prio Protocol - 4 steps

### 4. Publish phase

- servers publish their individual accumulator values
- final aggregate is computed by summing accumulators
- final aggregate statistic obtained with AFE decoding:

$$A_1, A_2, \dots, A_s$$

$$Decode(\sigma) \in \mathbb{A}$$

$$\sigma = \sum_{j=1}^s A_j = \sum_{i=1}^n Trunc_{k'} y_i$$

# Protocol Security Properties

- **robustness** against malicious clients holds if:
  - SNIP construction is sound - malicious client submissions are detected via SNIPs
- **f-privacy**, only the final aggregate statistic is revealed, holds if:
  - one server is honest
  - AFE is f-private
  - SNIP is zero-knowledge
- **anonymity** holds if:
  - function  $f$  is symmetric - the order of inputs does not affect the output

$$f(x_1, \dots, x_{n-m}) = f(x'_1, \dots, x'_{n-m})$$

$$(x'_1, \dots, x'_{n-m}) = \text{SORT}(x_1, \dots, x_{n-m})$$

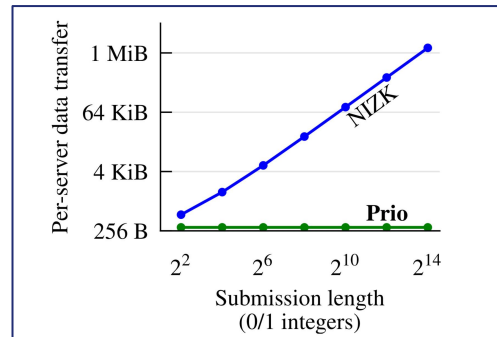
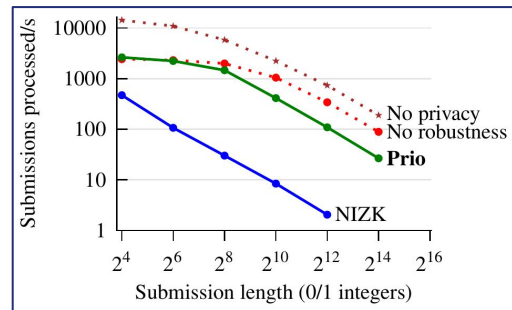


## 8. Evaluation

- Prio Client performance:
  - $\sim 0.03$  sec for a 100-integer submission on a workstation;  
 $\sim 0.1$  sec on a smartphone (2010-12 hardware).
- Prio Server throughput:
  - Outperforms NIZK-based scheme by 10x on average
  - Adding more server does not significantly affect throughput

Field size	Workstation		Smartphone	
	87-bit	265-bit	87-bit	265-bit
Multipl. in field ( $\mu$ s)	1.013	1.485	11.218	14.930
<b>L = 10</b>	0.003	0.004	0.017	0.024
<b>L = 100</b>	0.024	0.035	0.110	0.167
<b>L = 1000</b>	0.214	0.334	1.028	2.102

Time in seconds for a client to generate a Prio submission of L four-bit integers



## 9. Discussion - Limitations

- **Selective Denial-of-Service Attack**
- Intersection Attack
- *Robustness against faulty servers*
  - May be implemented but lowers the privacy guarantees
  - robust against  $k$  faulty servers (out of  $s$ )  $\Rightarrow$  protects privacy against at most  $s-k-1$  malicious servers

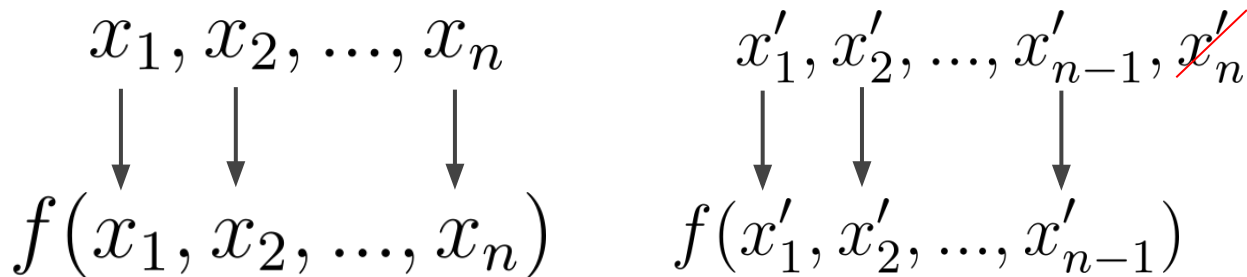
~~$x_1$~~ ,  ~~$x_2$~~ , ...,  $x_i$ , ...,  ~~$x_n$~~



$f(x_{\text{honest}}, x_{\text{evil}_1}, \dots, x_{\text{evil}_m})$

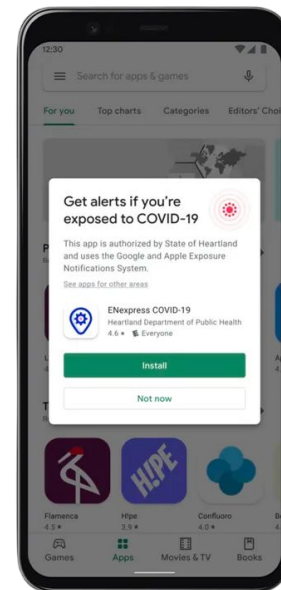
## 9. Discussion - Limitations

- Selective Denial-of-Service Attack
- **Intersection Attack**
- *Robustness against faulty servers*
  - May be implemented but lowers the privacy guarantees
  - robust against  $k$  faulty servers (out of  $s$ )  $\Rightarrow$  protects privacy against at most  $s-k-1$  malicious servers



## 9. Discussion - Deployments

- Many large-scale deployments since paper publication.
- During the *COVID-19* pandemic, *Apple* and *Google* introduced *Exposure Notification Privacy-preserving Analytics* to alert users about potential contact with individuals infected.
- Based on *Prio*.
- No one could access information about who received notifications or the identities of contacts.
- Aggregated insight were sent to *public health agencies*.



# Conclusion

- Prio allows the aggregation of complexe statistics on private client data.
- Uses additive secret-sharing, SNIP and AFEs.
- More efficient and scalable than traditional protocols.
- Has many practical applications.

Thank you for your attention !

# Appendix

# Detailed computation for $\sigma$

Define the following values, where  $s$  is a constant representing the number of servers:

$$\begin{aligned}x &= \sum_i [x]_i & a &= \sum_i [a]_i \\f(r) &= \sum_i [f(r)]_i & b &= \sum_i [b]_i \\r \cdot g(r) &= \sum_i [r \cdot g(r)]_i & c &= \sum_i [c]_i \\h(M) &= \sum_i [h]_i(M) & d &= f(r) - a \\& & e &= r \cdot g(r) - b\end{aligned}$$

$$\begin{aligned}\sigma &= \sum_i (de/s + d[b]_i + e[a]_i + [c]_i - [r \cdot h(r)]_i) \\&= de + db + ea + c - r \cdot h(r) \\&= (f(r) - a)(r \cdot g(r) - b) + (f(r) - a)b + (r \cdot g(r) - b)a + c - r \cdot h(r) \\&= r \cdot (f(r)g(r) - h(r)) + (c - ab)\end{aligned}$$