

Secure and Fault-Tolerant Aggregation for FL

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Overview of Secure Aggregation

- Definition and Applications
- Threat Model
- Integration in Federated Learning

Fault-Tolerant Secure Aggregation

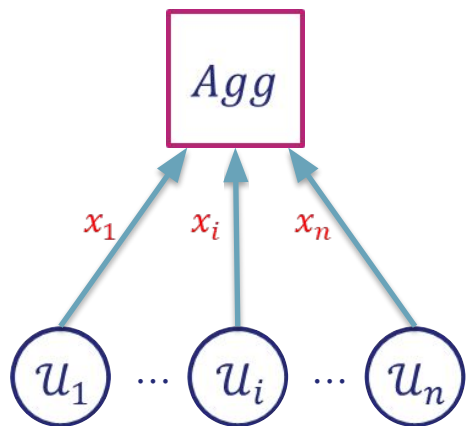
- Joye-Libert (JL) SA scheme
- User drop problem
- Our Threshold JL scheme
- Our protocol

Comparison with SOTA

Conclusion

What is Secure Aggregation?

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



■ x_i is a private user input

■ How to compute X without trusting Agg ?

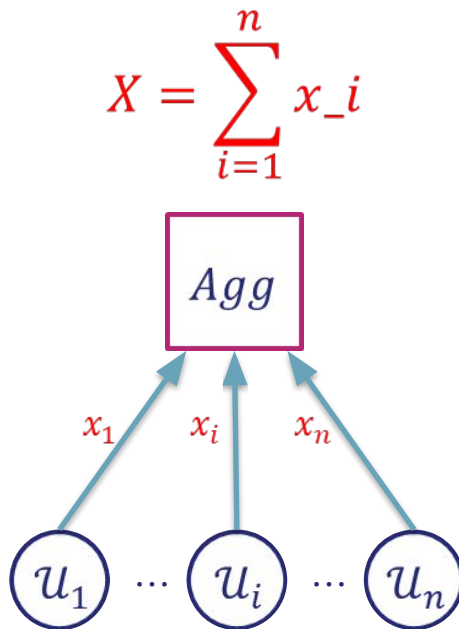
Secure Aggregation Threat Model

Threat Model

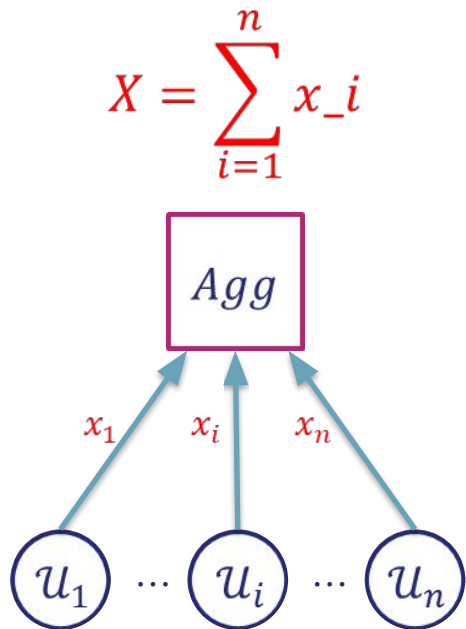
- **Malicious Aggregator colluding with users**

Security Requirement:

- **Aggregator Obliviousness** : The aggregator cannot learn more than the **sum** of the honest users inputs



Application of Secure Aggregation



■ Voting Systems: $x_i \in \{0, 1\}$

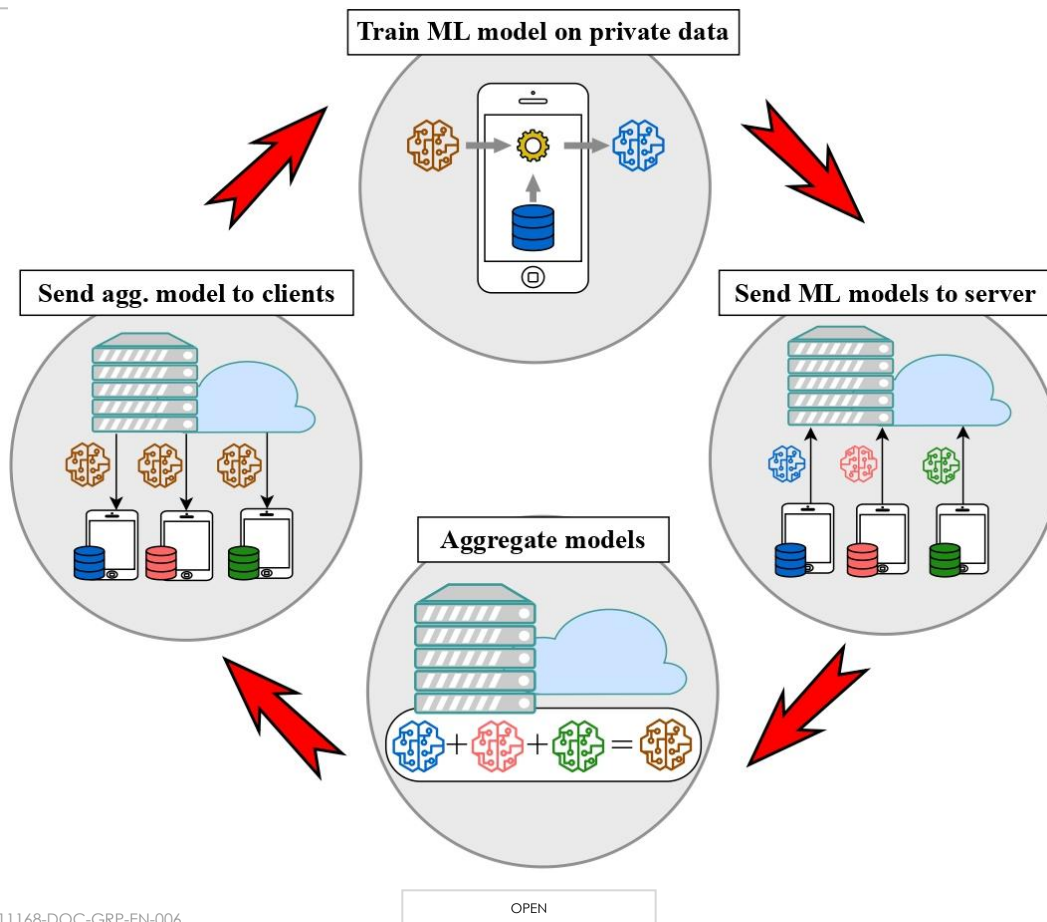
■ Rating Systems: $x_i \in \{0, 1, \dots, 10\}$

■ Smart Meters: $x_i \in [0, N_{max}]$

■ Federated Learning: $x_i \in [0, N_{max}]^m$

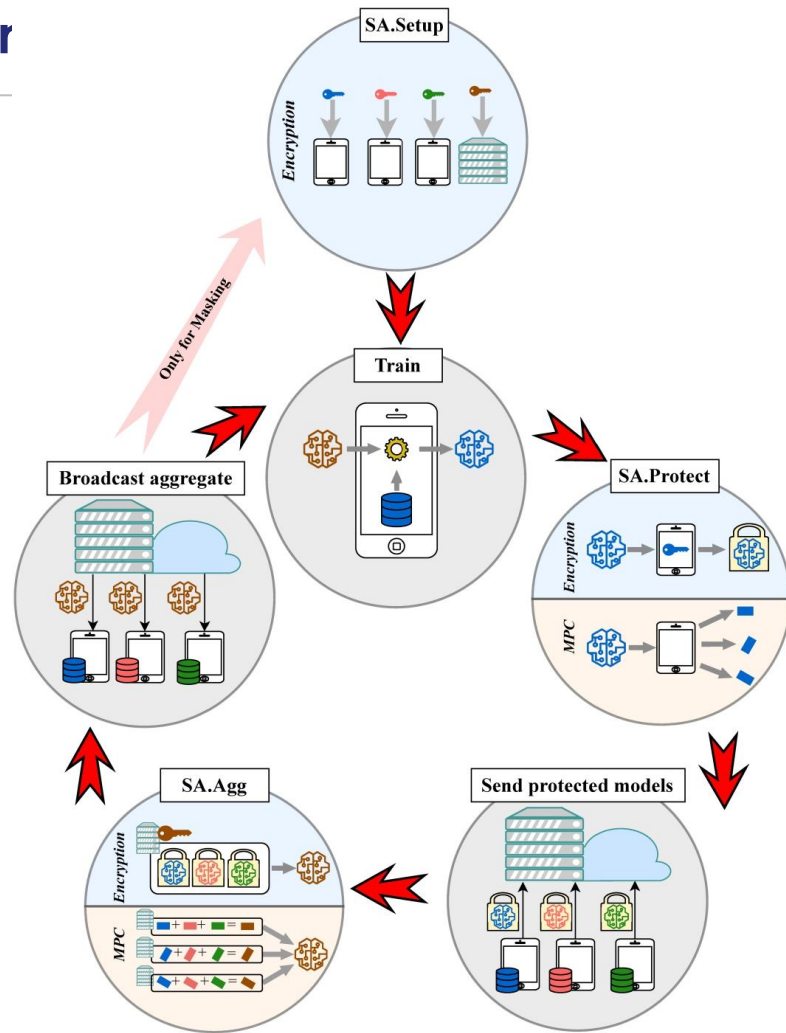
Federated Learning

What is FL?



Inference Attacks

- Leak information about the private data set from the trained model
- Membership Inference Attacks
- Reconstruction Attacks
- Data Properties Inference Attacks



AHE-Based Secure Aggregation

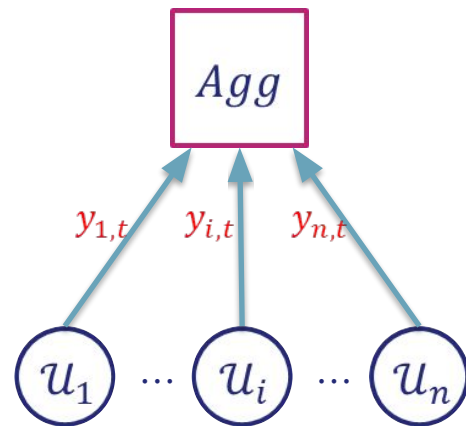
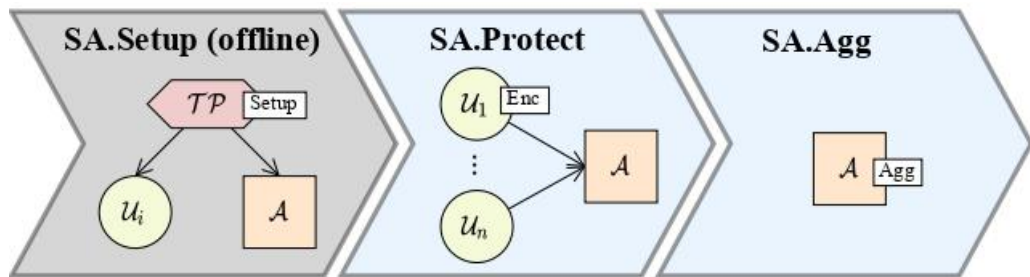
Joye-Libert Scheme

➤ **Setup**(λ): generate modulus N , hash H and key k_i for each user s.t. $\sum_i k_i = -k_a$

➤ **Protect**($pp, k_i, t, x_{i,t}$): $y_{i,t} = (1 + x_{i,t}N)H(t)^{k_i} \bmod N^2$

➤ **Agg**($pp, k_a, \{y_{i,t}\}_{\forall i}$): $X = \frac{H(t)^{k_a} \prod_i y_{i,t} - 1}{N}$

$$X = \text{Agg}(pp, k_a, \{y_{1,t}, \dots, y_{n,t}\})$$



Protect($pp, k_i, t, x_{i,t}$)

Threshold Joye-Libert Secure Aggregation

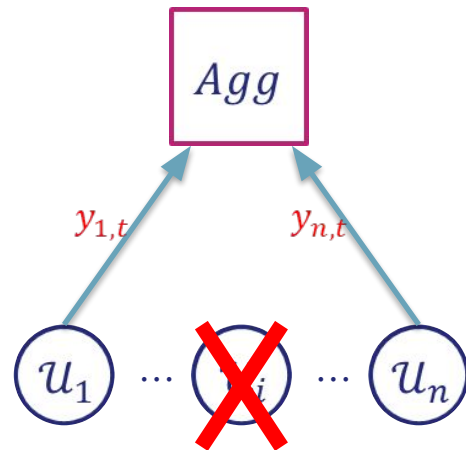
Problem of clients dropout

JL scheme is not fault-tolerant

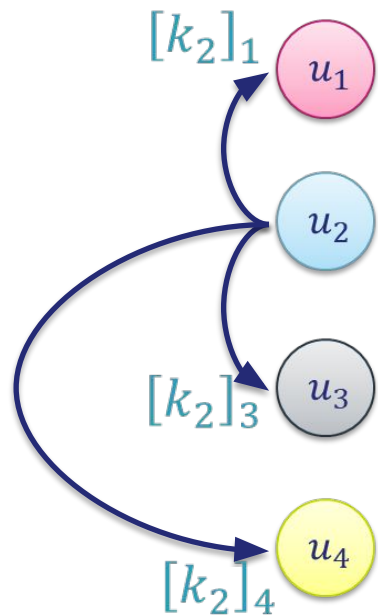
- If one or more clients dropped: $\sum_i k_i \neq -k_a$
- Agg cannot be computed

Threshold Joye-Libert scheme (TJL):

- Using Shamir's secret sharing,
 - Each client shares its protection key k_i
- When client u_i drops:
 - Any t -out-of- n clients encrypt on behalf u_i using the shares of k_i
- The server aggregate the protected inputs
- Minimum number of honest users $t > \frac{2n}{3}$ honest users

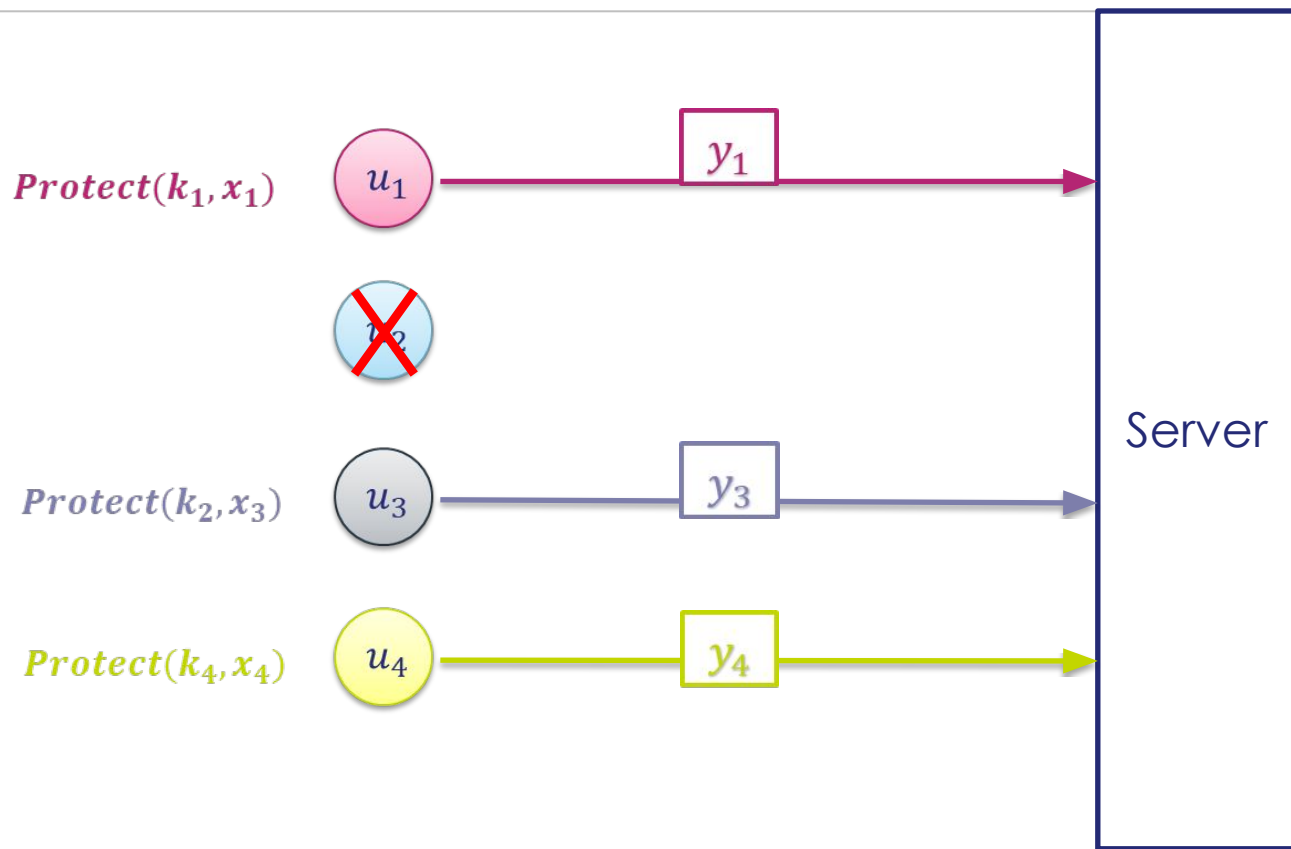


Fault Tolerant Secure Aggregation- Setup

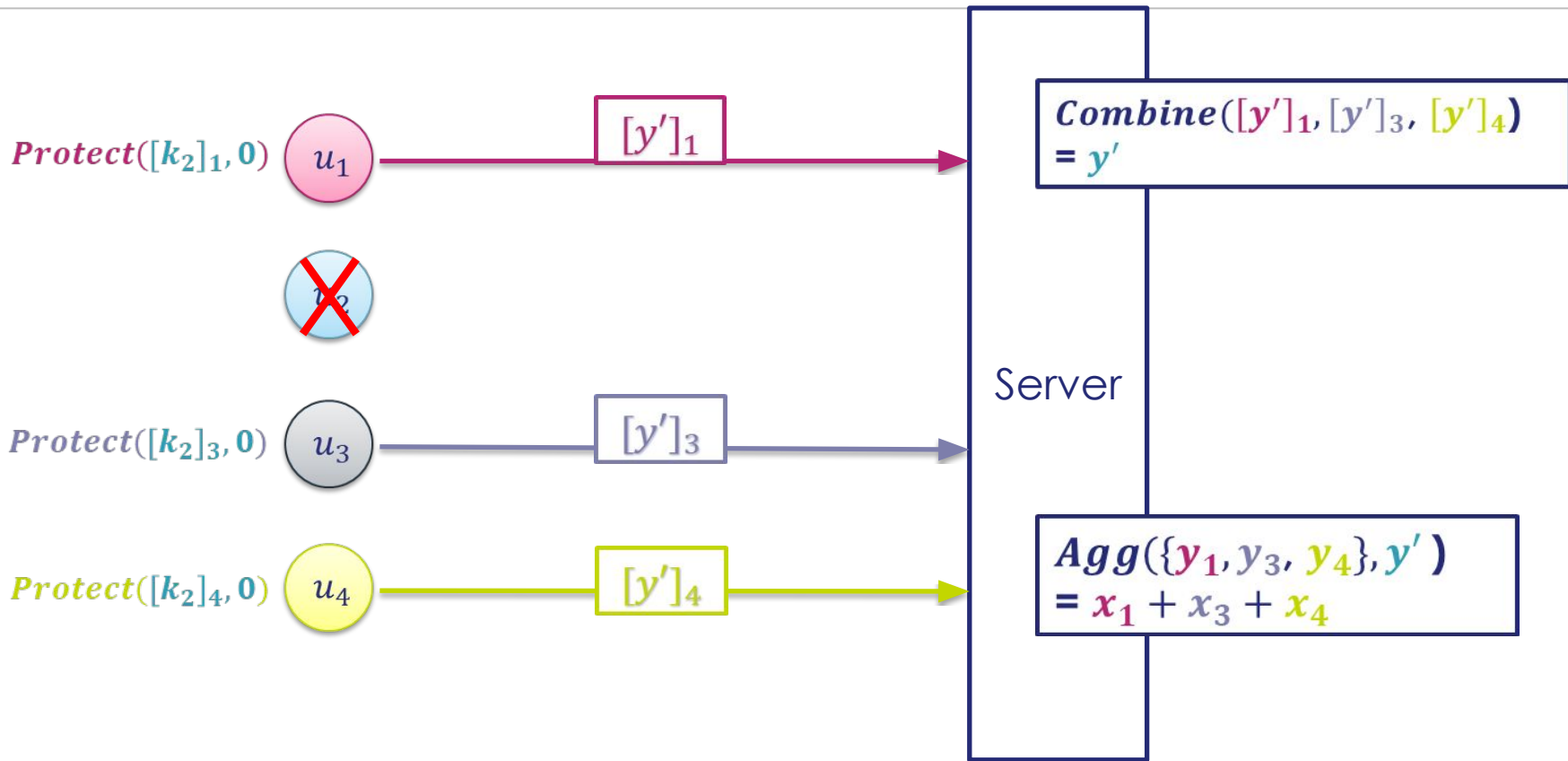


Server

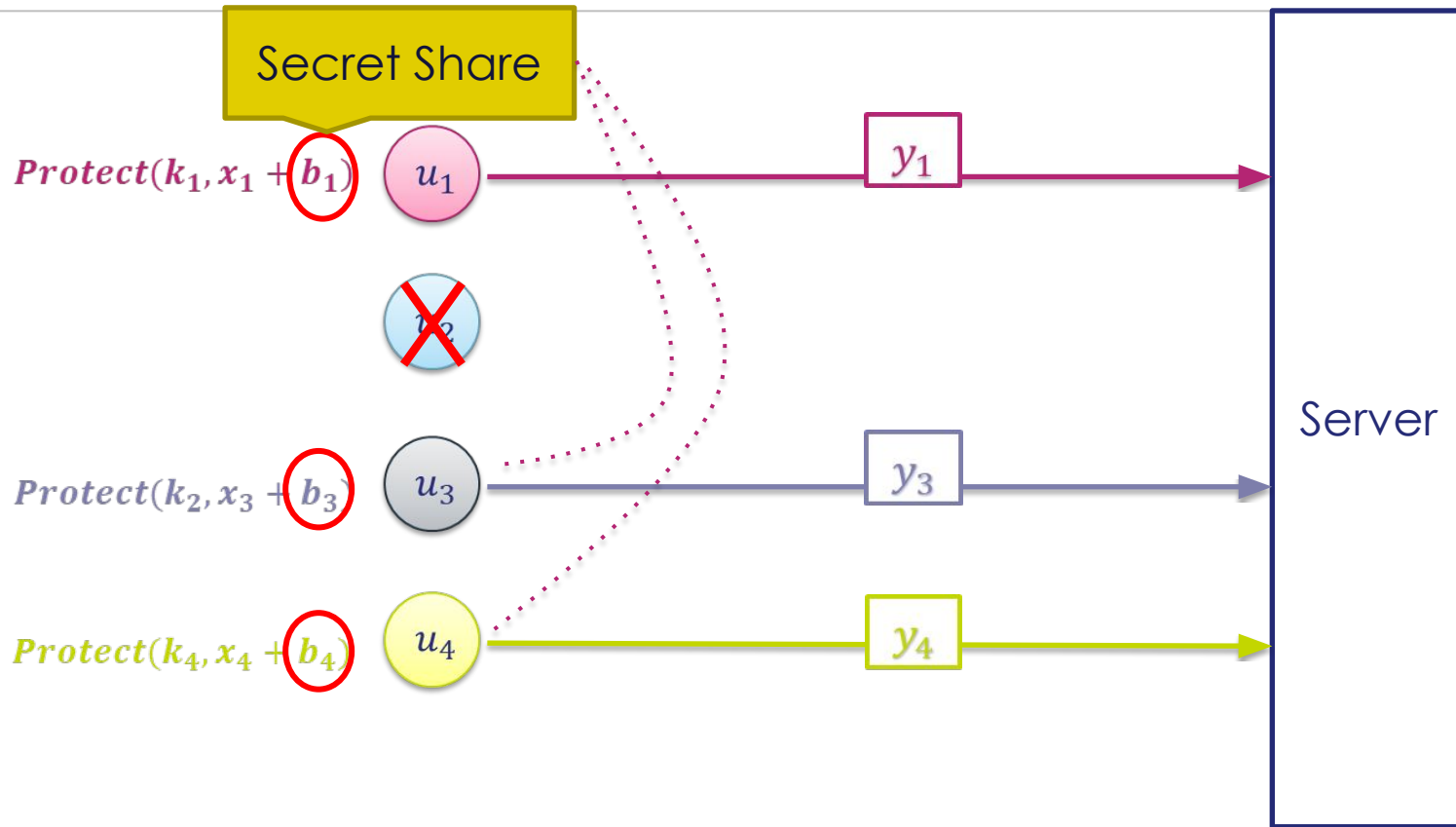
Fault Tolerant Secure Aggregation- Online - Encrypt



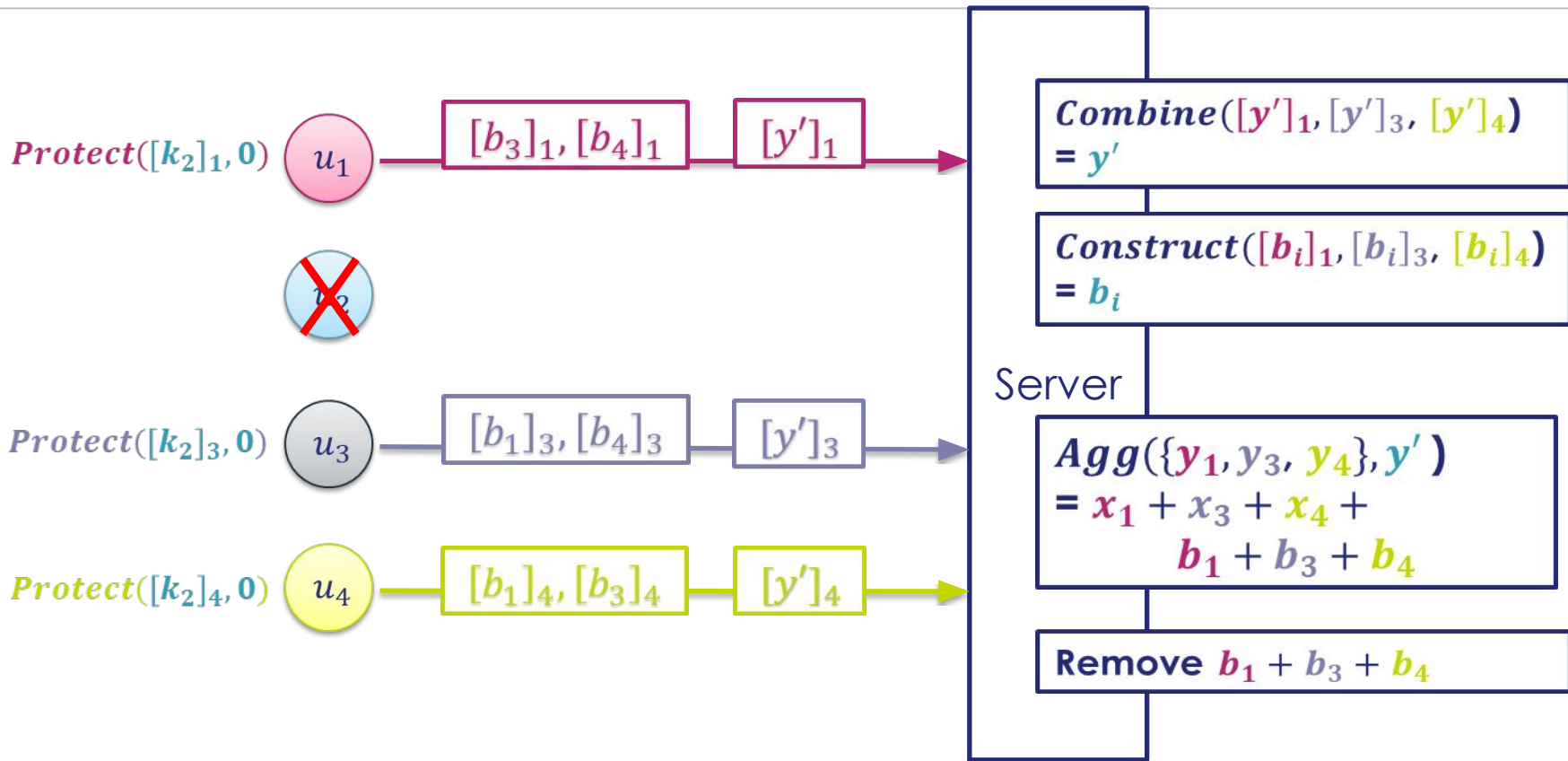
Fault Tolerant Secure Aggregation- Online - Construct



Fault Tolerant Secure Aggregation- Online - Encrypt



Fault Tolerant Secure Aggregation- Online - Construct



Our Solution vs [2] : Client computation

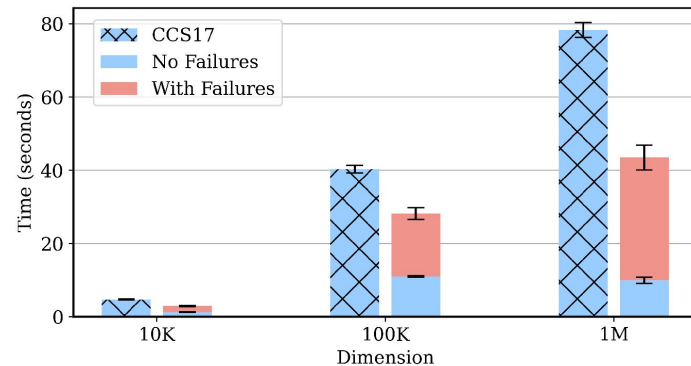
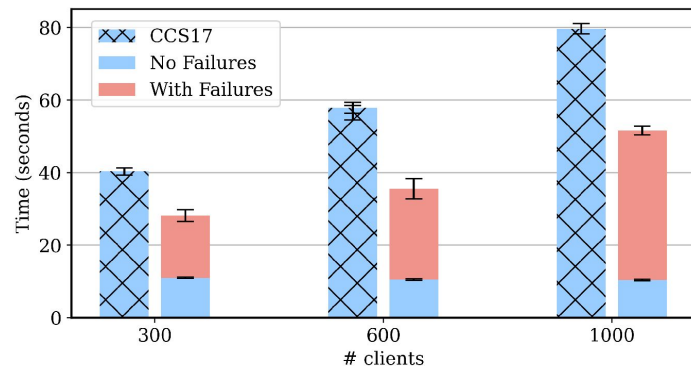
CCS17 [2]

Our Solution

Better scalability of the client's computation

- Our solution is $\times 2.7$ faster with 100 clients and $\times 4.6$ faster with 600

Same scalability w.r.t input dimension (m)



Conclusion and Future Work

Conclusion

- We presented TJL scheme
- We presented FTSA (first Fault-Tolerant SA based on AHE)
- We compared to the SoTA (better scalability for the clients)

Future Work

- Optimizing the scalability
- Achieving Aggregate Unforgeability

Thanks!

Fault Tolerant Secure Aggregation- Online - Construct

$\text{Protect}([k_2]_1 + [k_4]_1, 0)$

u_1

$[y']_1$

~~u_2~~

$\text{Protect}([k_2]_3 + [k_4]_3, 0)$

u_3

$[y']_3$

~~u_4~~

$\text{Combine}([y']_1, [y']_3)$
 $= y'$

Server

$\text{Agg}(\{y_1, y_3\}, y')$
 $= x_1 + x_3$

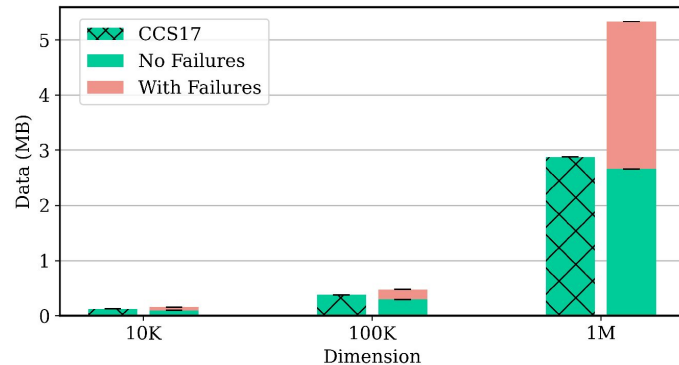
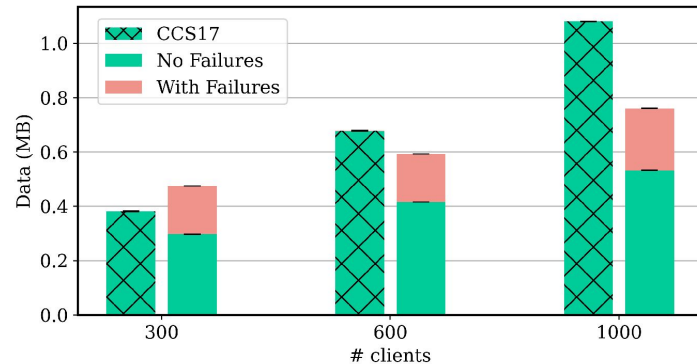
Our Solution vs [2] : Client communication

CCS17 [2]

Our Solution

Better scalability w.r.t # clients (n)

Worst scalability w.r.t input dimension (m)



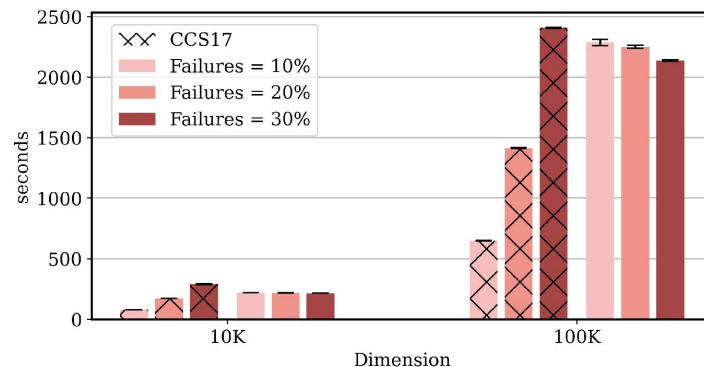
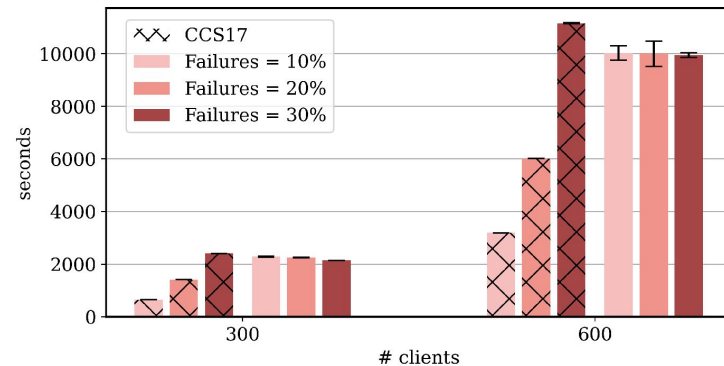
Our Solution vs [2]: Server computation

CCS17 [2]

Our Solution

Better scalability w.r.t the number of dropped clients (d)

➤ Better performance on high 30% dropouts



Fault Tolerant Secure Aggregation using AHE

Threshold Joye-Libert scheme (TJL)

- $(k_a, \{k_i\}_{\forall i \in [n]}, pp) \leftarrow Setup(\lambda)$: generates keys and public parameters
- $\{j, [k_i]_j\} \leftarrow SKShare(k_i, t, n)$: share the secret key of user \mathcal{U}_1
- $[y'_t]_i \leftarrow ShareProtect(pp, \{[k_j]_i\}_{\forall u \in U'}, t)$: protect a zero-value using the shares of **dropped** clients
- $y'_t \leftarrow ShareCombine(pp, \{[y'_t]_i\}_{\forall i \in U'}, t)$: combine the protected zero-values shares of **dropped** clients
- $y_{i,t} \leftarrow Protect(pp, k_i, t, x_{i,t})$: protect an input $x_{i,t}$ of **online** client
- $\sum_{\forall i} x_{i,t} \leftarrow Agg(pp, k_a, \{y_{i,t}\}_{\forall i \in U'}, y'_\tau)$: aggregate all protected input of **online** clients and zero-values of **dropped** ones

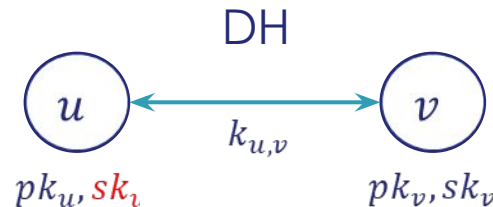
Fault Tolerant Secure Aggregation

State of the Art: Fault Tolerant Masking [2]

- Use DH key agreement between two clients to agree on masking keys k_u

$$k_u = \sum_{u < v} \text{PRG}(k_{u,v}) - \sum_{u > v} \text{PRG}(k_{u,v})$$

- Notice that: $\sum_u k_u = 0$



- *Protect*: $y_u = x_u + k_u + \text{PRG}(b_u) \bmod R$
 - b_u is a random generated seed
 - To make the scheme Fault-Tolerant, use Shamir's secret sharing to share sk_u and b_u of each user
 - If user u is **online**: The server collects shares of b_u and reconstruct it
 - If user u **dropped**: The server collects shares of sk_u and reconstruct it
- It then re-computes all DH agreements and recover the mask k_u