

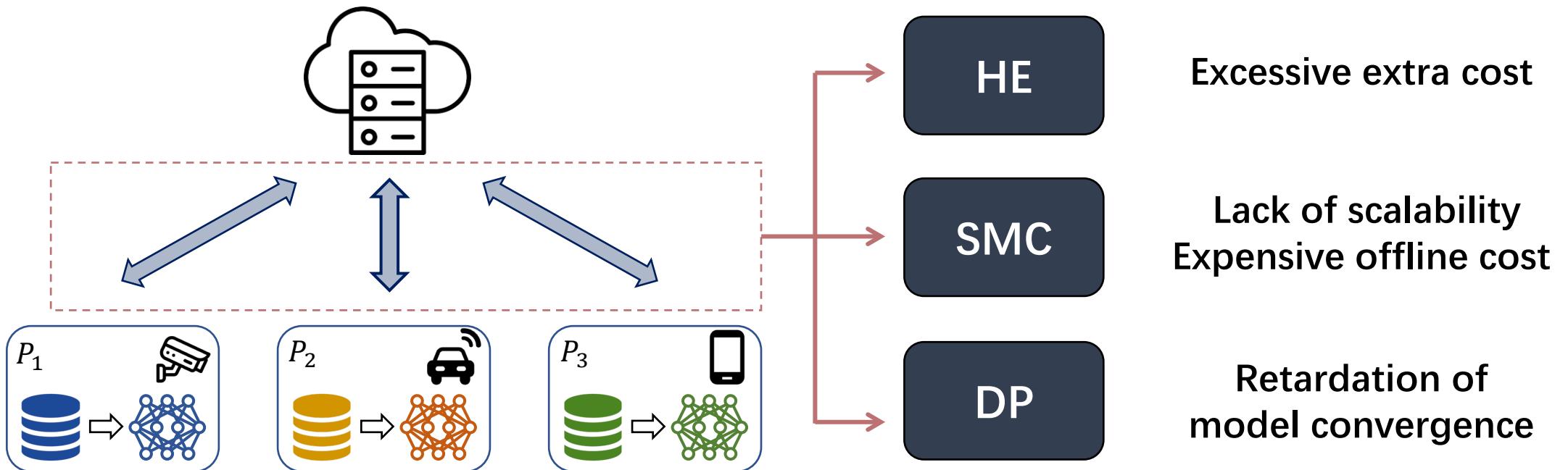
FLSwitch: Towards Secure and Fast Model Aggregation for Federated Deep Learning with a Learning State-Aware Switch

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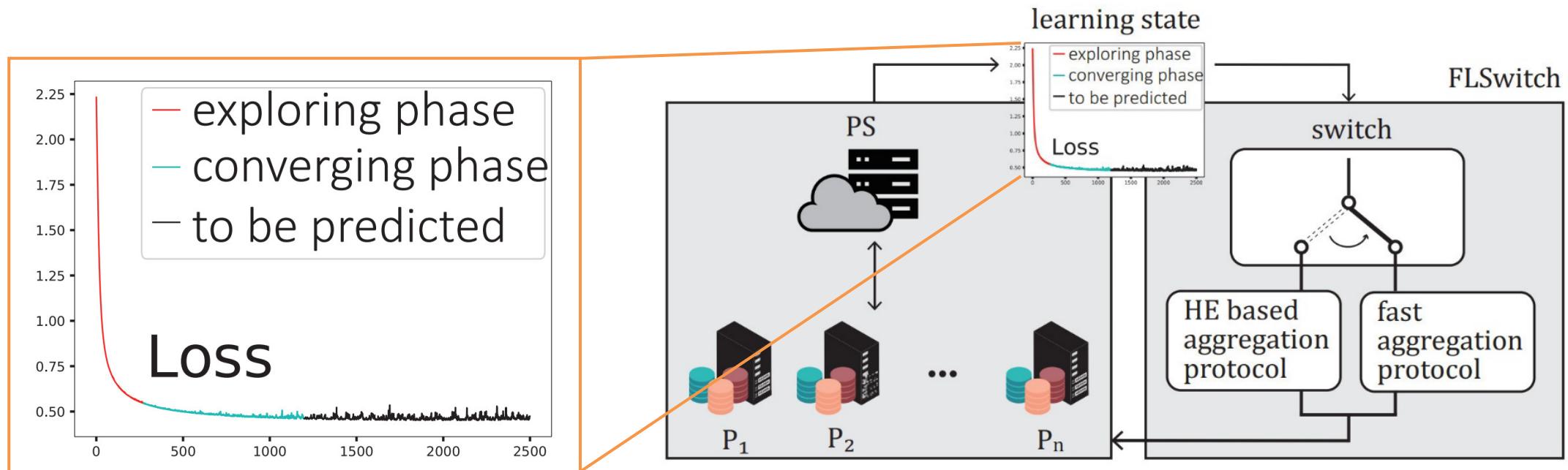
*ByteDance Ltd. Shenzhen, China

Related Work



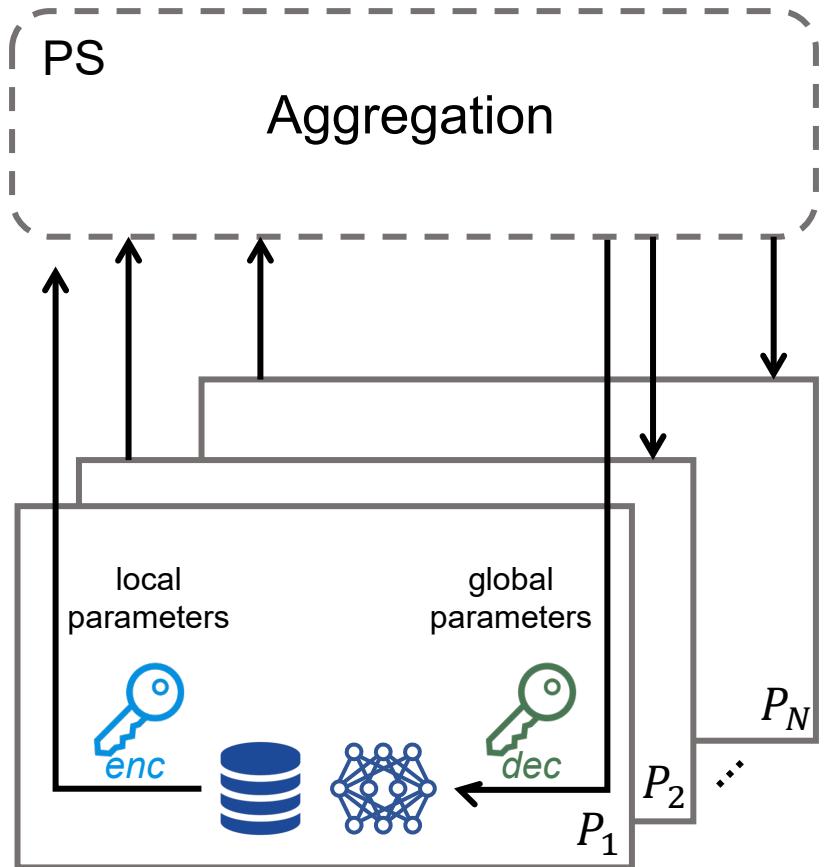
FLSwitch

- Overview



Secure Model Aggregation (SMA)

Threat Model



- semi-honest PS and participants
- secure communication channel
- no collusion allowed between the adversarial PS and any participant
- collusion of up to $N - 2$ adversarial participants is allowed

Our Work

- We propose a residual encoding based HE protocol (RBPHE), outperforming the existing solutions in single instruction multiple data operation (SIMD).
- We propose a fast SMA protocol by utilizing FL characteristics and lightweight cryptographical tools for further efficiency improvement, which significantly speeds up conventional SMA designs.
- We design a switch model based on meta-learning, monitoring FL tasks and switching between protocols dynamically.

RBPHE

- RBPHE provides a flexible encoding range extension method.
- RBPHE reduces the precision loss of updated parameters via dynamic range extension.
- RBPHE achieves more efficient batching and takes less amortized overhead for critical operations.

RBPHE

- Comparison of HE-based SMA solutions

Input Size	Scheme*	Size (KB)	Enc (ms)	Dec (ms)	Add (ms)	Mul (ms)
4096	CKKS (16bit)	1280.1	16	6	0.6	1
	BatchCrypt (8bit)	16.6	350	105	1	/
	BatchCrypt (16bit)	25.7	528	157	2	/
	RBPHE (8bit)	19.1	261	79	1	0.8
	RBPHE (16bit)	30.3	411	122	2	1
65536	CKKS (16bit)	10241.1	127	44	5	8
	BatchCrypt (8bit)	256.8	5442	1633	24	/
	BatchCrypt (16bit)	403.3	8275	2452	36	/
	RBPHE (8bit)	297.8	4142	1254	17	13
	RBPHE (16bit)	479.9	6497	1916	28	21

* We use the implementation of CKKS in the SEAL library. The implementation of BatchCrypt and RBPHE is based on python-paillier.

Anchor negotiation

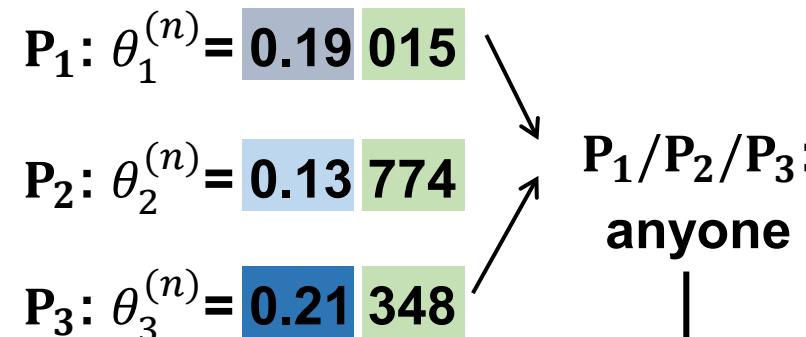
- parameter split

$$\theta = a \cdot 10^{-\gamma} + r$$

anchor residue

split in $\gamma = 3$
0.81975e-1 { anchor: 0.81
residue: 0.000975

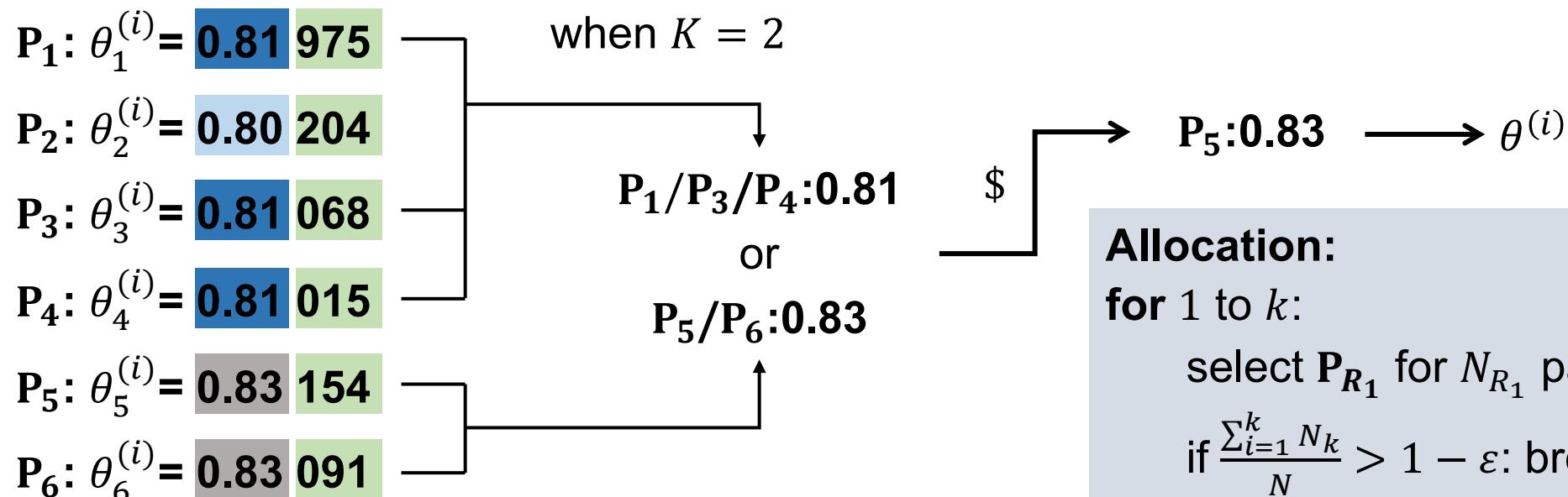
- select representatives



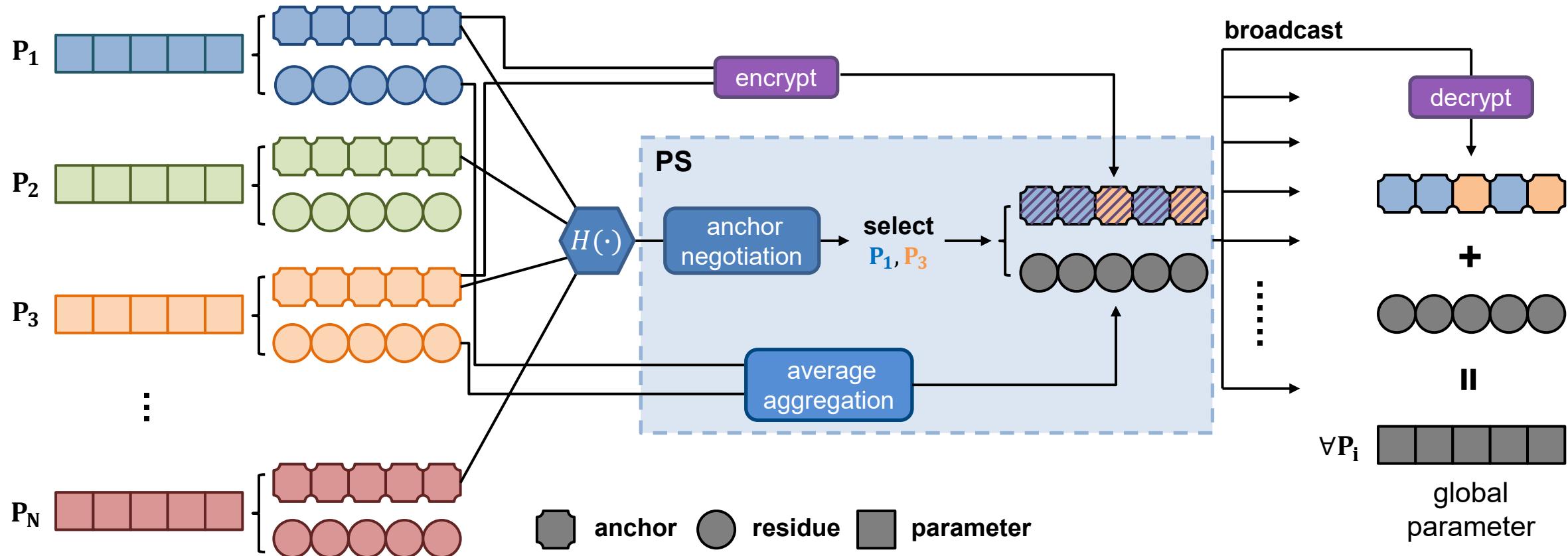
aggregate
 θ
global
parameter

Anchor negotiation

- anchorK



Fast SMA



Analysis of Fast SMA

- **cost analysis**

Computation: $N\text{enc}(\theta) \rightarrow K\text{enc}(\theta) + O(|\theta|N\log K)$

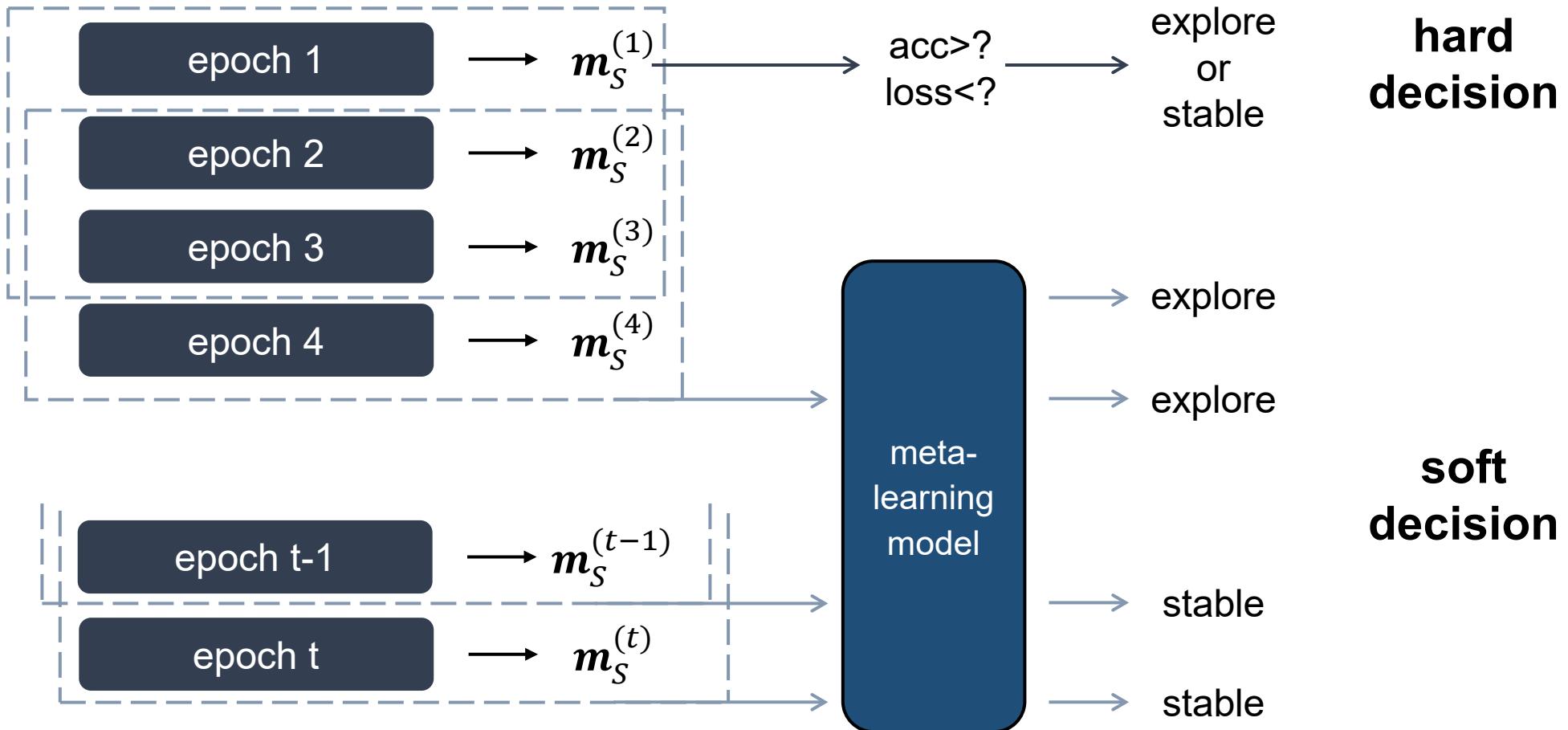
Communication: $N|\mathbf{c}| \rightarrow K|\mathbf{c}| + K|\mathbf{r}|$

- **security analysis**

Anchors are protected via HE

Meaningless hijacked residues without anchors

Learning State-Aware Switch



Experiment Setup

- **Dataset**

MNIST, FASHION-MNIST, CIFAR10 and CIFAR100

- **Model**

For MNIST family: 3-layer fully-connected NN

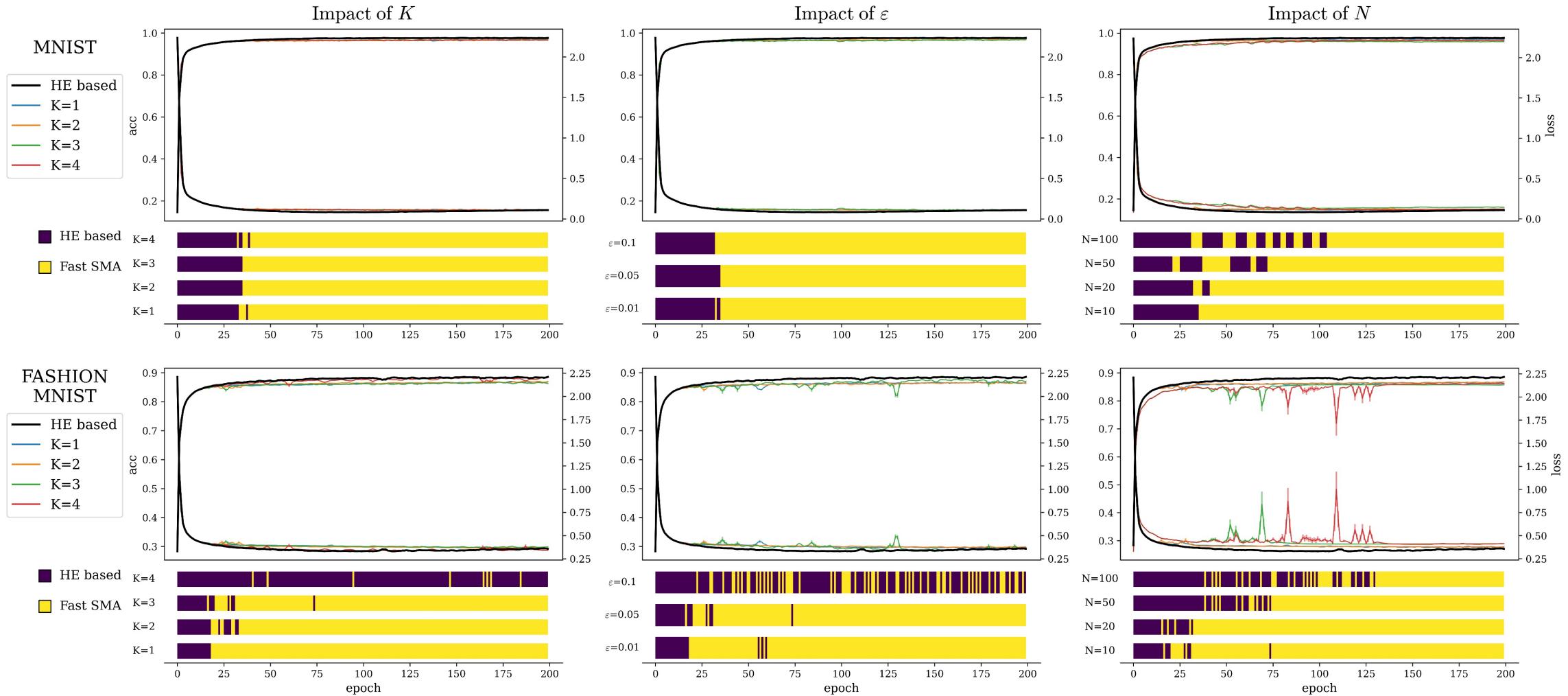
For CIFAR family: 20-layer ResNet

- **HE algorithms**

Paillier, Batchcrypt, CKKS and our RBPHE

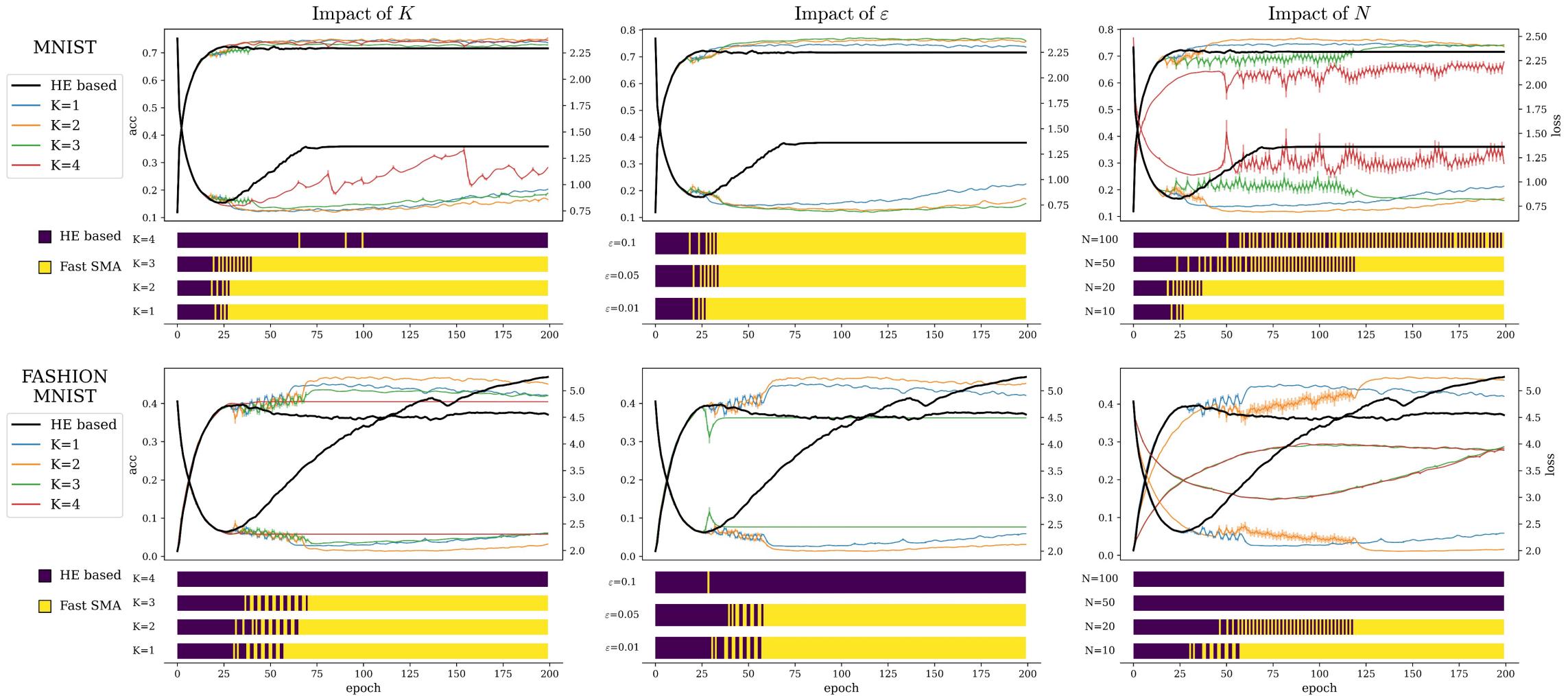
Experiments

- **MNIST**



Experiments

- **CIFAR**



Experiments

Execution time

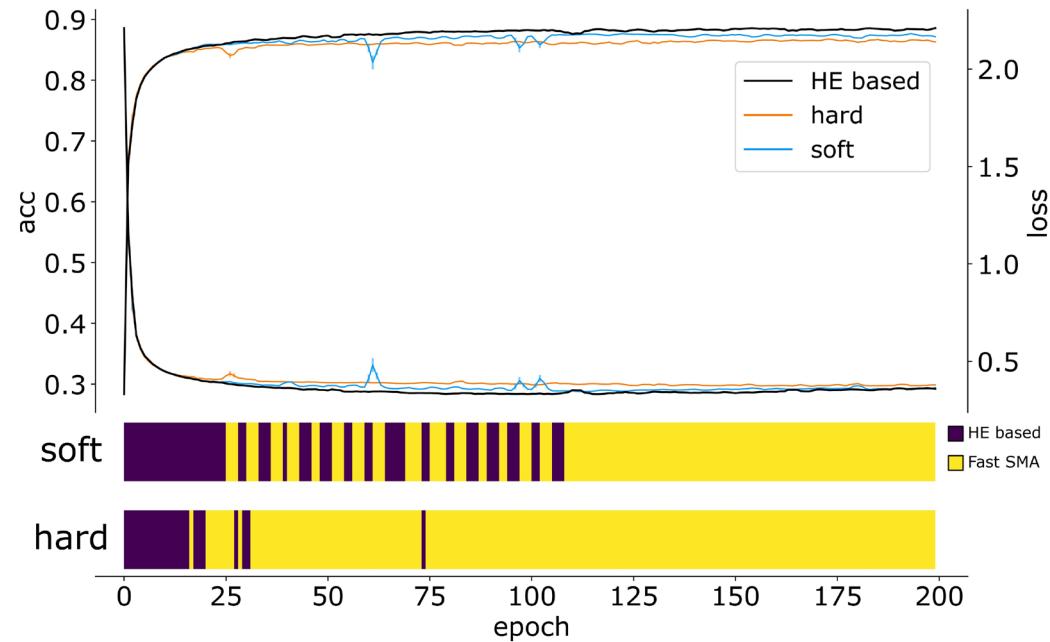
Dataset	Protocol	Clients10		Clients50		Clients100	
		Client	Server	Client	Server	Client	Server
FASHION MNIST	FastAgg	2.22	2.36	5.03	13.48	28.21	3.79
	RBPHE	8.64	0.19	23.68	1.04	47.15	2.08
	Paillier	8.72	0.07	25.53	0.41	52.69	0.86
	Batchcrypt	9.64	0.26	26.24	1.48	51.83	3.06
CIFAR10	FastAgg	24.92	17.91	56.83	26.38	165.74	19.23
	RBPHE	43.17	0.95	122.21	5.12	233.03	10.28
	Paillier	47.10	0.35	130.46	2.24	263.89	4.45
	Batchcrypt	48.59	1.20	131.27	6.65	266.74	14.98

Communication cost

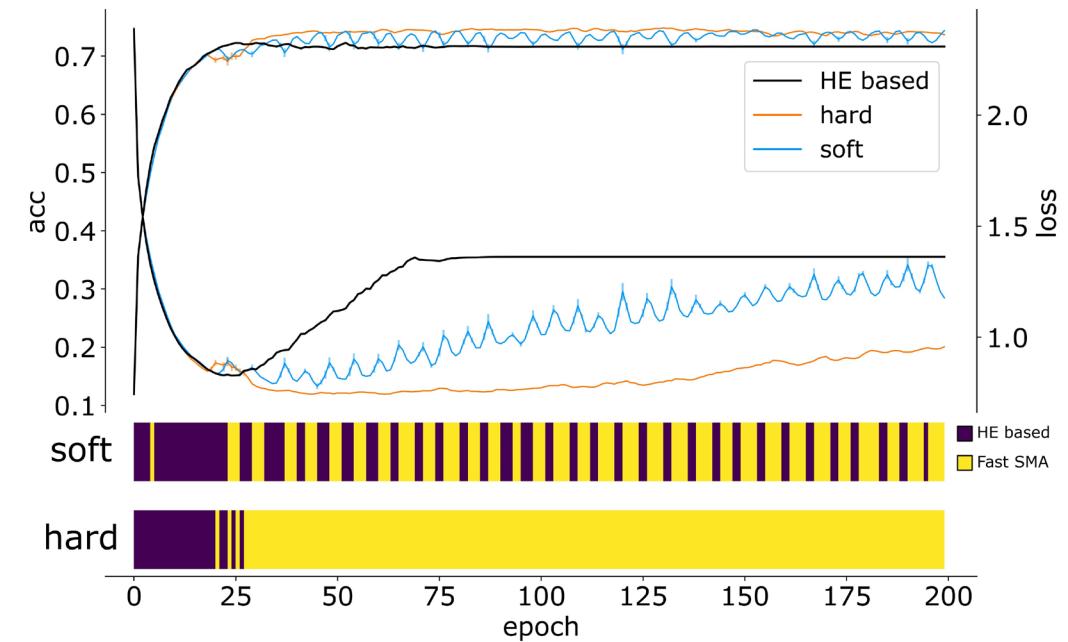
Dataset	Protocol	Clients10		Clients50		Clients100	
		Client	Server	Client	Server	Client	Server
FASHION MNIST	FastAgg	0.64	7.88	0.53	33.34	0.48	59.59
	RBPHE	0.37	3.65	0.37	18.26	0.37	36.51
	Paillier	0.28	2.82	0.31	15.56	0.32	32.22
	Batchcrypt	0.29	2.94	0.29	14.68	0.29	29.39
CIFAR10	FastAgg	4.02	43.79	2.62	163.82	2.21	260.33
	RBPHE	1.80	18.01	1.80	90.07	1.80	180.1
	Paillier	1.52	15.23	1.68	83.82	1.74	173.68
	Batchcrypt	1.44	14.49	1.44	72.46	1.44	144.86

Experiments

- State-aware switch model



FASHION-MNIST



CIFAR10

Thanks!

question?