Efficient 3PC for Binary Circuits with Application to Maliciously-Secure DNN Inference

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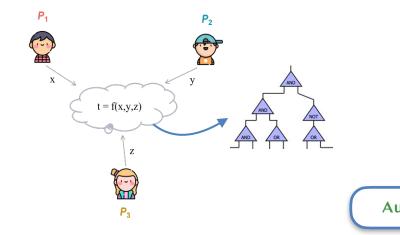






Problem Setting

Three-Party Computation (on Binary Circuits)





Previous Works

Cut-and-choose + Beaver Triples

- Use cut-and-choose to generate valid Beaver triples and use them to check online AND triples
- Low computational cost
- High communication complexity
 - <u>9 bits</u> per AND gate per party [FLNW17]
 - <u>7 *bits*</u> per AND gate per party [ABFL+17]

Distributed Zero-knowledge Proofs

- Use distributed zero-knowledge proofs to verify local semi-honest AND gate computations
- High computational cost
 - rely on *binary extension fields* [BGIN19]
- Low communication complexity
 - 1 bit per AND gate per party (amortized)

Question

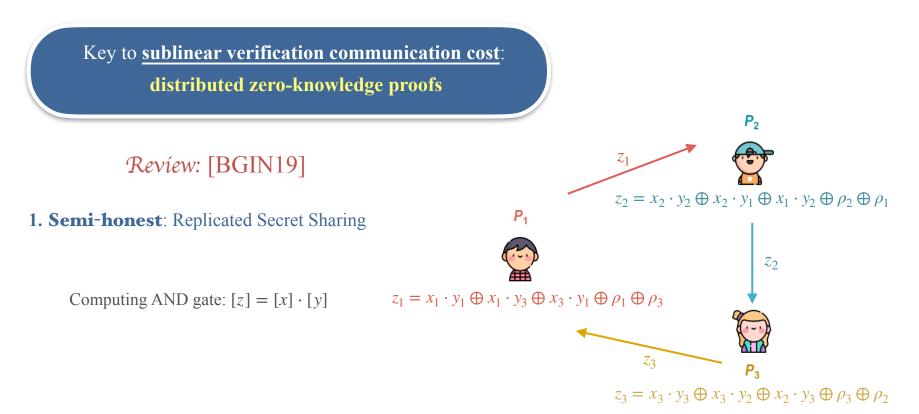
Can we achieve the same communication complexity as [BGIN19] and also a comparable concrete efficiency as [FLNW17, ABFL+17] ?

Our Results

- A Maliciously-Secure 3PC Protocol for Binary Circuits
 - Communication complexity
 - the same as [BGIN19]
 - $9 \times$ lower than [FLNW17], $7 \times$ lower than [ABFL+17]
 - Computational cost
 - comparable with [FLNW17]
 - 3.5× faster than [BGIN19]

- Application to DNN Inference
 - SqueezeNet
 - DenseNet
 - ResNet50

Starting Point: [BGIN19]



Starting Point: [BGIN19]

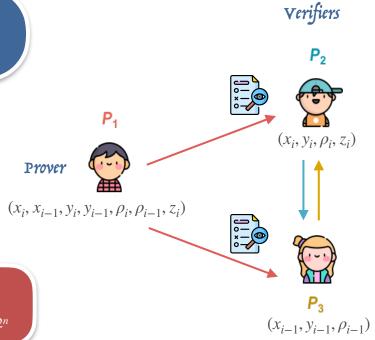
Key to sublinear verification communication cost: distributed zero-knowledge proofs [BBCG+19]

Review: [BGIN19]

2. Verification: Verify semi-honest AND computations

 $z_i = x_i \cdot y_i \oplus x_i \cdot y_{i-1} \oplus x_{i-1} \cdot y_i \oplus \rho_i \oplus \rho_{i-1}$

Lift \mathbb{F}_2 computations up to its extension field \mathbb{F}_{2^n} to apply distributed zero-knowledge proofs over \mathbb{F}_{2^n}



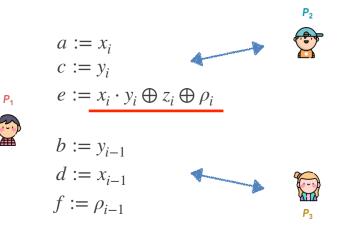
Lift \mathbb{F}_2 computations up to prime fields \mathbb{F}_p , and exploit the **algebraic structure** of prime fields to further reduce the computational cost of distributed zkps

1. Reduce the relation

$$z_{i} = x_{i} \cdot y_{i} \oplus x_{i} \cdot y_{i-1} \oplus x_{i-1} \cdot y_{i} \oplus \rho_{i} \oplus \rho_{i-1}$$

$$\downarrow$$

$$a \cdot b \oplus c \cdot d \oplus e \oplus f = 0$$



Lift \mathbb{F}_2 computations up to prime fields \mathbb{F}_p , and exploit the algebraic structure of prime fields to further reduce the computational cost of dzkps

2. Transform to Prime Fields

$$\alpha \oplus \beta = \alpha + \beta - 2\alpha\beta \mod p = \alpha(1 - 2\beta) + \beta \mod p$$

$$a \cdot b \oplus c \cdot d \oplus e \oplus f = -2(a \cdot c \cdot (1 - 2e)) \cdot (b \cdot d \cdot (1 - 2f)) + (c \cdot (1 - 2e)) \cdot (d \cdot (1 - 2f)) + (a \cdot (1 - 2e)) \cdot (b \cdot (1 - 2f)) - \frac{1}{2}((1 - 2e) \cdot (1 - 2f)) + \frac{1}{2} \mod p$$

Lift \mathbb{F}_2 computations up to prime fields \mathbb{F}_p , and exploit the algebraic structure of prime fields to further reduce the computational cost of dzkps

2. Transform to Prime Fields

 P_2

 P_3

$$a \cdot b \oplus c \cdot d \oplus e \oplus f = \cdots \cdots$$

 $\sum_{k=1}^4 g_k \cdot h_k + 1/2 = 0 \mod p$

 $g_{1} := -2a \cdot c \cdot (1 - 2e) \mod p$ $g_{2} := c \cdot (1 - 2e) \mod p$ $g_{3} := a \cdot (1 - 2e) \mod p$ $g_{4} := -(1 - 2e)/2 \mod p$

 $h_1 := b \cdot d \cdot (1 - 2f) \mod p$ $h_2 := d \cdot (1 - 2f) \mod p$ $h_3 := b \cdot (1 - 2f) \mod p$ $h_4 := 1 - 2f \mod p$

 P_1

Lift \mathbb{F}_2 computations up to prime fields \mathbb{F}_p , and exploit the algebraic structure of prime fields to further reduce the computational cost of dzkps

3. Batch Verifying AND Gates

For some Incorrect AND triple:

$$\sum_{k=1}^{4} g_k \cdot h_k + 1/2 = a \cdot b \oplus c \cdot d \oplus e \oplus f = 1$$

Batch checking *m* **AND triples:**

 $\sum_{\ell=1}^{m} \left(\sum_{k=1}^{4} g_{k}^{(\ell)} \cdot h_{k}^{(\ell)} + 1/2\right) = 0 \quad \text{all AND triples are correct (given } m < p)$ No need for random linear combination

Our Techniques: Optimizations

Lift \mathbb{F}_2 computations up to prime fields \mathbb{F}_p , and exploit the algebraic structure of prime fields to further reduce the computational cost of dzkps

Use the Powers of 2 as Coefficients for Linear Combination

Batch checking m AND triples:Enable to use **native CPU computations** for batch computing
 \mathbb{F}_p data directly from \mathbb{F}_2 data (without first transforming them to \mathbb{F}_p) $\sum_{\ell=1}^{m} (\sum_{k=1}^{4} g_k^{(\ell)} \cdot h_k^{(\ell)} + 1/2) = 0$ $\sum_{i=1}^{m} (\sum_{j=1}^{2(i-1) \mod 32} \sum_{j=1}^{k} g_k^{(i)} \cdot h_k^{(i)} + 1/2) = 0$ Choosing $p = 2^{61} - 1$ is sufficient for any $m \le 2^{33}$

Our Techniques: Optimizations

Lift \mathbb{F}_2 computations up to prime fields \mathbb{F}_p , and exploit the algebraic structure of prime fields to further reduce the computational cost of dzkps

Other Optimizations

Lookup Table for Polynomial Evaluation

Use Mersenne Fields for Fast Arithmetic Computation

Speed up Inner-product Operations

Performance

Performance on Pure Binary Circuits

Depth		Semi	Ours	BGIN19	FLNW17
1	LAN Time	0.12	1.15	3.42	0.95
	WAN Time	2.97	3.31	5.67	11.74
10	LAN Time	0.12	1.14	3.78	0.90
	WAN Time	2.18	2.58	4.61	11.41
100	LAN Time	0.12	1.14	3.84	0.91
	WAN Time	5.18	5.85	6.78	14.94
1000	LAN Time	0.18	1.16	3.87	0.96
	WAN Time	41.05	41.83	42.76	51.05
10000	LAN Time	0.70	1.36	4.05	1.50
	WAN Time	401.60	402.47	403.35	412.35
	Comm.	24.00	24.80	24.57	224.16

Comparison with [BGIN19]

- Communication: almost the same
- End-to-end time: $3 \sim 3.4 \times$ faster in LAN, $1 \sim 1.8 \times$ faster in WAN

Table 2: Time (s), communication (MB) for computing circuits of 64 million AND gates with different depths.

Performance

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Comparison with [FLNW17]

- **Communication**: 9× lower
- End-to-end time: $1 \sim 1.3 \times$ slower in LAN, $2.6 \sim 4.2 \times$ faster in WAN

Table 2: Time (s), communication (MB) for computing circuits of 64 million AND gates with different depths.

Performance

Performance on DNN Inference

Model	# of Threads		Semi+Semi	${\tt SW}+{\tt Semi}^1$	SW + Ours	SW + BGIN19	SW + FLNW17
ResNet-50	1	LAN Time	89.68	336.71	372.97	582.55	391.23
	32	LAN Time	18.66	56.64	63.33	81.48	87.48
	32	WAN Time	544.15	1969.48	2048.89	2096.17	2786.22
		Comm.	7537.86	27791.9	27846.10	27830.40	41114.30
DenseNet	1	LAN Time	63.42	305.89	375.72	622.83	371.124
	32	LAN Time	12.98	57.07	66.13	84.60	94.39
	32	WAN Time	713.42	1994.98	2070.69	2096.17	2842.19
		Comm.	8919.85	31924.50	31993.40	31973.5	48709.60
SqueezeNet	1	LAN Time	13.61	49.13	58.89	106.20	63.09
	32	LAN Time	2.23	9.93	11.28	14.17	15.70
	32	WAN Time	200.19	432.05	448.96	455.29	674.26
		Comm.	1403.22	4803.36	4816.58	4812.76	8047.35

Time cost for achieving malicious security for the binary part

• SW+Ours is 15% \sim 33% of SW+BGIN19 in LAN, and 7% \sim 10% of SW+FLNW17 in WAN

An Extra Finding

A hidden security issue in probabilistic truncation protocols

prepare
$$[r], [r/2^d]$$

 \downarrow
compute and open $[x + r] = [x] + [r]$
 \downarrow
output $(x + r)/2^d - [r/2^d]$

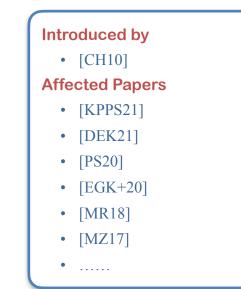
The same randomness is used for both
 protecting the privacy of the secret value, and
 sampling the 1-bit rounding error probabilistically

An Extra Finding

A hidden security issue in probabilistic truncation protocols

prepare
$$[r], [r/2^d]$$

 \downarrow
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 \downarrow
output $(x + r)/2^d - [r/2^d]$



Thanks !

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