Communication and Randomness Lower Bounds for Secure Multiparty Computation B. Tech Project

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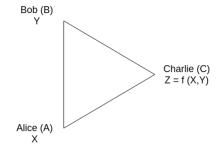
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Three-Party Secure Computation - Introduction

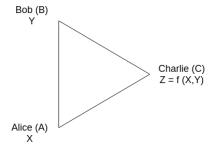
- ▶ Alice (A) and Bob (B) have private data X and Y respectively
- ▶ Charlie (C) computes a function Z = f(X, Y)
- ► There is a private channel between every pair of parties



Three-Party Secure Computation Model

Three-Party Secure Computation - Objectives

- Charlie must compute Z with zero probability of error
- ▶ Charlie must not learn X and Y, more than what Z reveals
- ▶ Alice must not learn Y and Z, more than what X reveals
- ▶ Bob must not learn X and Z, more than what Y reveals

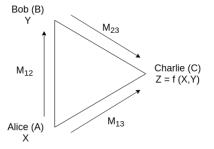


Three-Party Secure Computation Model

Three-Party Secure Computation - One-shot FKN Protocol

- 1. Alice chooses $M_{12} \in \mathcal{M}_{12}$, sends it to Bob privately
- 2. Alice sends M_{13} , a function of (M_{12}, X) to Charlie
- 3. Bob sends M_{23} , a function of (M_{12}, Y) to Charlie
- 4. Charlie computes \hat{Z} (estimate of Z) as a function of M_{13} , M_{23}

Find minimum $H(M_{12}), H(M_{23}), H(M_{13})$ for this to succeed.



Messages sent in one-shot protocol

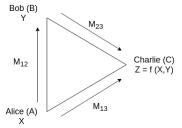
Three-Party Secure Computation - Formal Goals

Secrecy:

- ▶ Alice chooses M_{12} independent of X (secrecy for Alice and Bob)
- ▶ $(M_{13}, M_{23}) Z (X, Y)$ is a Markov Chain (secrecy for Charlie) or $Pr(m_{13}, m_{23}|x, y) = Pr(m_{13}, m_{23}|x', y')$ if f(x, y) = f(x', y')

Correctness:

▶ supp $(M_{13}, M_{23}|x, y) \cap \text{supp}(M_{13}, M_{23}|x'y') = \Phi$ if $f(x, y) \neq f(x', y')$

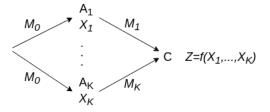


Three-Party Secure Computation Model

Multiparty Secure Computation - Extended FKN Protocol

Parties A_1, \ldots, A_k with inputs X_1, \ldots, X_k

- 1. Common randomness $M_0 \in \mathcal{M}_0$ given to parties A_1, \dots, A_k
- 2. Each party sends M_i , a function of (M_0, X_i) to Charlie
- 3. Charlie computes \hat{Z} as a function of (M_1, \ldots, M_k)



Multiparty Secure Computation Model

Applications of Multi-Party Secure Computation

► Secure Audio Teleconferencing:

- Bridge mediates communication by detecting which party's signal has the maximum amplitude
- Compute max of encrypted signals without decrypting them

Secure auctions:

- N parties bid their price for a product
- Each party's bid to be unknown to other parties and to the seller
- Seller must correctly determine the highest bidder

► Benchmark Analysis:

- ► A third party agent compares performance of difference companies based on certain parameters
- Performance parameters of each company should not be leaked to other companies

Machine Learning:

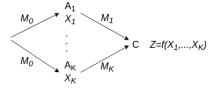
- Organization collects data from several users
- ► Each user's private data must not be learnt by other users or by the organization
- Accurately train machine learning models on the data

Literature Study - Existence of Secure Protocols

- ▶ Feige, Kilian, Naor (1994) proved that secure computation protocol exists for any function $f: \{0,1\}^n \times \{0,1\}^n \to \{0,1\}$
- Amount of communication and randomness may be exponential in input length
- Communication and randomness polynomial if f is in nondeterministic logspace
- Protocol for secure multiparty computation of logical AND

Secure Computation of AND - Multiparty Protocol

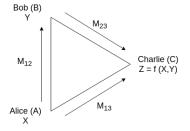
- $lacksquare X_1,\ldots,X_k\in\{0,1\}$ and $f(X_1,\ldots,X_k)=X_1\wedge\ldots\wedge X_k$
- Protocol by Feige, Kilian and Naor where:
 - ▶ Choose a prime p > k
 - ▶ M_0 is 0 < r < p and r_1, \ldots, r_k with $\sum_{i=1}^k r_i = 0 \mod p$
 - ▶ Each party sends $M_i = r(1 x_i) + r_i \mod p$ if its input is x_i
 - ► Charlie outputs Z = 1 iff $\sum_{i=1}^{k} M_i = 0 \mod p$
- ▶ Cardinality of $|\mathcal{M}_0| = p^{k-1}(p-1)^k$ and $|\mathcal{M}_i| = p$



One-shot protocol for multiparty AND

Secure Computation of AND - Three-party Protocol

- ▶ $X, Y \in \{0,1\}$ and $f(X, Y) = X \land Y$
- Protocol by Feige, Kilian and Naor where:
 - ▶ M_{12} is a random permutation of (0,1,2), say (α,β,γ)
 - ▶ Alice sends $M_{13} = \alpha$ if X = 1 and β if X = 0
 - ▶ Bob sends $M_{23} = \alpha$ if Y = 1 and γ if Y = 0
 - ▶ Charlie computes Z = 1 if $M_{13} = M_{23}$, and Z = 0 otherwise
- ▶ Cardinality of $|\mathcal{M}_{12}|$ = 6 and $|\mathcal{M}_{13}|$ = $|\mathcal{M}_{23}|$ = 3



One-shot protocol for three-party AND

Literature Study - Existing Bounds

 Information theoretic bounds for general functions, using interactive protocols with private randomness. For AND,

$$H(M_{13}) \ge \log_2 3$$
, $H(M_{23}) \ge \log_2 3$ and $H(M_{12}) \ge 1.826$

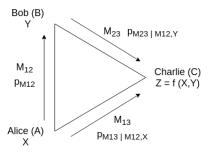
- ► Given protocol : $H(M_{13}), H(M_{23}) = \log_2 3$ and $H(M_{12}) = \log_2 6$
- Common randomness lower bound not achieved by the protocol
- For one-shot protocols without private randomness (e.g. FKN),

$$H(M_{13}) \ge \log_2 3$$
, $H(M_{23}) \ge \log_2 3$ and $H(M_{12}) \ge \log_2 6$

▶ The FKN protocol for AND is optimal for $H(M_{13})$ but may not be for $H(M_{12})$ over more general protocols

Private Randomness

- ▶ M_{13} may not a deterministic function of (M_{12}, X) .
- ► M_{13} drawn from a distribution $p_{M_{13}|M_{12},X}$
- ▶ There exist functions f(X, Y) where using private randomness decreases the lower bound on $H(M_{12})$
- Can we trade off common randomness for private randomness?



Messages and their distributions

Private Randomness - Example of Protocol

Consider the following function with $X, Y \sim \text{Unif}\{0, 1, 2\}$:

$$f(X,Y) = \begin{cases} 2 & \text{if } X = 2 \text{ or } Y = 2 \\ X \oplus Y & \text{otherwise} \end{cases}$$

Protocol with k' and k'' drawn from Alice's and Bob's private randomness

- 1. M_{12} contains a uniformly chosen permutation of (0,1,2), say (α,β,γ) and a uniform bit k.
- 2. Alice sends $M_{13}=(\alpha,X\oplus k)$ if $X\in\{0,1\}$, and (β,k') if X=2.
- 3. Bob sends $M_{23}=(\alpha,Y\oplus k)$ if $Y\in\{0,1\}$, and (γ,k'') if Y=2.
- 4. Charlie finds Z = 2 if $M_{13}(1) = M_{23}(1)$, and $Z = M_{13}(2) \oplus M_{23}(2)$ otherwise.

Private Randomness - Example of Protocol

Consider the following function with $X, Y \sim \text{Unif}\{0, 1, 2\}$:

$$f(X,Y) = \begin{cases} 2 & \text{if } X = 2 \text{ or } Y = 2 \\ X \oplus Y & \text{otherwise} \end{cases}$$

Theoretical bounds without private randomness :

$$H(M_{13}) \ge 2.3137$$
, $H(M_{23}) \ge 2.3137$ and $H(M_{12}) \ge 3.8987$

Achieved by the protocol:

$$H(M_{13}) = H(M_{23}) = \log_2 3 + 1 \approx 2.5850$$

 $H(M_{12}) = \log_2 6 + 1 \approx 3.5850$

My Work - Proof under Private Randomness

- ▶ $|\mathcal{M}_{12}| \ge 6$ holds for non-interactive protocols with private randomness for secure computation of AND
- Using two properties of the AND function:

1. $f(1,1) \neq f(1,0)$. So correctness requires that

- $supp((M_{13}, M_{23})|XY = 10) \cap supp((M_{13}, M_{23})|XY = 11) = \Phi$
- 2. f(0,0) = f(0,1) = f(1,0). So secrecy requires that $\sup((M_{13}, M_{23})|XY = 00) = \sup((M_{13}, M_{23})|XY = 01) = \sup((M_{13}, M_{23})|XY = 10)$
- ▶ Proof that any $|\mathcal{M}_{12}|$ < 6 cannot satisfy these two conditions

Proof - Private Randomness by One Party

- Suppose only Bob uses private randomness.
- ightharpoonup Fix $M_{12}=m_{12}$ and let $a=m_{13}(m_{12},0)$. Let

$$b \in \text{supp}(M_{23}|Y=0, M_{12}=m_{12}),$$

 $b' \in \text{supp}(M_{23}|Y=1, M_{12}=m_{12})$
 $\implies [a, b] \in \text{supp}((M_{13}, M_{23})|XY=00)$

From the condition for secrecy,

$$[a, b'] \in \text{supp}((M_{13}, M_{23})|XY = 00)$$

▶ For every a in supp $(M_{13}|X=0)$, there exists $m_{12}, m'_{12} \in \mathcal{M}_{12}$ such that $a=m_{13}(m_{12},0)=m_{13}(m'_{12},0)$

$$|\mathcal{M}_{12}| \ge 2|\mathsf{supp}(M_{13}|X=0)| = 2|\mathsf{supp}(M_{13})| \ge 6$$

Proof - Private Randomness by Both Parties

- Each row represents an input combination
- Each column represents one value of the common randomness
- ▶ Each cell in the table shows $supp(M_{13}, M_{23}|X, Y, M_{12})$
- $\begin{array}{l} \blacktriangleright \ A_1^0 = supp(M_{13}|X=0,M_{12}=1), \\ B_1^0 = supp(M_{23}|Y=0,M_{12}=1) \end{array}$
- $A_1^0 B_1^0$ denotes set product of A_1^0 and B_1^0

	$M_{12} = 1$	$M_{12} = 2$	$M_{12}=3$	$M_{12} = 4$	$M_{12} = 5$	$M_{12} = 6$
XY = 00	$A_1^0 B_1^0$	$A_2^0 B_2^0$	$A_3^0 B_3^0$	$A_4^0 B_4^0$	$A_5^0 B_5^0$	$A_6^0 B_6^0$
XY = 01	$A_1^0 B_1^1$	$A_2^0 B_2^1$	$A_3^0 B_3^1$	$A_4^0 B_4^1$	$A_5^0 B_5^1$	$A_6^0 B_6^1$
XY = 10	$A_1^1 B_1^0$	$A_2^1 B_2^0$	$A_3^1 B_3^0$	$A_4^1 B_4^0$	$A_5^1 B_5^0$	$A_6^1 B_6^0$
XY = 11	$A_1^1 B_1^1$	$A_2^1 B_2^1$	$A_3^1 B_3^1$	$A_4^1 B_4^1$	$A_5^1 B_5^1$	$A_6^1 B_6^1$

Proof - Private Randomness by Both Parties

1. $f(1,1) \neq f(1,0)$. Correctness requires that $supp((M_{13},M_{23})|XY=10) \cap supp((M_{13},M_{23})|XY=11) = \Phi$ Sets in row 4 must be disjoint with the sets in rows 1, 2 or 3

2.
$$f(0,0) = f(0,1) = f(1,0)$$
. Secrecy requires that $supp((M_{13}, M_{23})|XY = 00) = supp((M_{13}, M_{23})|XY = 01) = supp((M_{13}, M_{23})|XY = 10)$

The same sets must appear in rows 1, 2 and 3

<u>Proof</u> - Private Randomness by Both Parties

- ▶ For secrecy, we need at least $|\mathcal{M}_{12}| \ge 2$
- ▶ For $|\mathcal{M}_{12}| = 2$, we need $A_1^0 B_1^1 \subseteq A_2^0 B_2^0$ and $A_1^1 B_1^0 \subseteq A_2^0 B_2^0$
- $A_1^0, A_1^1 \subseteq A_2^0$ and $B_1^1, B_1^0 \subseteq B_2^0 \implies A_1^1 B_1^1 \subseteq A_2^0 B_2^0$ (Contradiction)

	$M_{12} = 1$	$M_{12} = 2$	$M_{12} = 3$	$M_{12} = 4$	$M_{12} = 5$	$M_{12} = 6$
XY = 00	$A_1^0 B_1^0$	$ \begin{array}{c} A_2^0 B_2^0 \\ (A_1^1 B_1^0) \\ (A_1^0 B_1^1) \end{array} $	$A_3^0 B_3^0$	$A_4^0 B_4^0$	$A_5^0 B_5^0$	$A_6^0 B_6^0$
XY = 00 $XY = 01$	$A_1^0 B_1^1$ $A_1^0 B_1^1$	$A_1^0 B_1^1$ $A_2^0 B_2^1$	$A_3^0 B_3^1$ $A_3^0 B_3^1$	$A_4^0 B_4^1$	$A_5^0 B_5^1$	$A_6^0 B_6^1$
<i>XY</i> = 10	$A_1^1 B_1^0$	$A_2^1 B_2^0$	$A_3^1 B_3^0$	$A_4^1 B_4^0$	$A_5^1 B_5^0$	$A_6^1 B_6^0$
XY = 11	$A_1^1 B_1^1$	$A_2^1 B_2^1$	$A_3^1 B_3^1$	$A_4^1 B_4^1$	$A_5^1 B_5^1$	$A_6^1 B_6^1$

<u>Proof</u> - Private Randomness by Both Parties

WLOG, assume

$$A_1^1 \cap A_2^0 \neq \Phi, \ B_1^0 \cap B_2^0 \neq \Phi, \ A_1^0 \cap A_3^0 \neq \Phi, \ B_1^1 \cap B_3^0 \neq \Phi$$
$$B_1^1 \cap B_2^0 = \Phi, \ B_1^1 \cap B_2^1 = \Phi, \ A_1^1 \cap A_3^0 = \Phi, \ A_1^1 \cap A_3^1 = \Phi$$

	$M_{12} = 1$	$M_{12} = 2$	$M_{12} = 3$	$M_{12} = 4$	$M_{12} = 5$	$M_{12} = 6$
<i>XY</i> = 00	$A_1^0 B_1^0$	$ \begin{array}{c} A_2^0 B_2^0 \\ (A_1^1 B_1^0) \end{array} $	$A_3^0 B_3^0 \\ (A_1^0 B_1^1)$	$A_4^0 B_4^0$	$A_5^0 B_5^0$	$A_6^0 B_6^0$
XY = 01	$A_1^0 B_1^1$	$A_2^0 B_2^1$	$A_3^0 B_3^1$	$A_4^0 B_4^1$	$A_5^0 B_5^1$	$A_6^0 B_6^1$
XY = 10	$A_1^1 B_1^0$	$A_2^1 B_2^0$	$A_3^1 B_3^0$	$A_4^1 B_4^0$	$A_5^1 B_5^0$	$A_6^1 B_6^0$
XY = 11	$A_1^1 B_1^1$	$A_2^1 B_2^1$	$A_3^1 B_3^1$	$A_4^1 B_4^1$	$A_5^1 B_5^1$	$A_6^1 B_6^1$

Proof - Private Randomness by Both Parties

- ▶ $A_1^1 \cap A_4^0 = A_1^1 \cap A_5^0 = \Phi$ or $A_1^1 \cap A_4^0 \neq \Phi, A_1^1 \cap A_5^0 \neq \Phi$ lead to a contradiction
- WLOG also assume

$$A_1^1 \cap A_4^0 \neq \Phi, \ A_1^1 \cap A_5^0 = \Phi, \ B_1^1 \cap B_4^0 = \Phi, \ B_1^1 \cap B_5^0 \neq \Phi$$

	$M_{12} = 1$	$M_{12} = 2$	$M_{12}=3$	$M_{12}=4$	$M_{12} = 5$	$M_{12} = 6$
<i>XY</i> = 00	$A_1^0 B_1^0$	$A_2^0 B_2^0 \\ (A_1^1 B_1^0)$	$A_3^0 B_3^0 \\ (A_1^0 B_1^1)$	$ \begin{array}{c} A_4^0 B_4^0 \\ (A_1^1 B_1^0) \end{array} $	$A_5^0 B_5^0 \\ (A_1^0 B_1^1)$	$A_6^0 B_6^0$
XY = 01	$A_1^0 B_1^1$	$A_2^0 B_2^1$	$A_3^0 B_3^1$	$A_4^0 B_4^1$	$A_5^0 B_5^1$	$A_6^0 B_6^1$
XY = 10	$A_1^1 B_1^0$	$A_2^1 B_2^0$	$A_3^1 B_3^0$	$A_4^1 B_4^0$	$A_5^1 B_5^0$	$A_6^1 B_6^0$
XY = 11	$A_1^1 B_1^1$	$A_2^1 B_2^1$	$A_3^1 B_3^1$	$A_4^1 B_4^1$	$A_5^1 B_5^1$	$A_6^1 B_6^1$

<u>Proof</u> - Private Randomness by Both Parties

- ► Enumerate all valid ways of splitting elements of $A_1^1 B_1^0$ in columns 2 and 4 and $A_1^0 B_1^1$ in columns 3 and 5
- For each case, assign other sets as per correctness and secrecy conditions
- Show that each of these cases leads to a contradiction

	$M_{12} = 1$	$M_{12} = 2$	$M_{12} = 3$	$M_{12} = 4$	$M_{12} = 5$	$M_{12} = 6$
XY = 00	$A_1^0 B_1^0$	$A_2^0 B_2^0 \\ (A_1^1 B_1^0)$	$A_3^0 B_3^0 \\ (A_1^0 B_1^1)$	$ \begin{array}{c} A_4^0 B_4^0 \\ (A_1^1 B_1^0) \end{array} $	$A_5^0 B_5^0 \\ (A_1^0 B_1^1)$	$A_6^0 B_6^0$
XY = 01	$A_1^0 B_1^1$	$A_2^0 B_2^1$	$A_3^0 B_3^1$	$A_4^0 B_4^1$	$A_5^0 B_5^1$	$A_6^0 B_6^1$
XY = 10	$A_1^1 B_1^0$	$A_2^1 B_2^0$	$A_3^1 B_3^0$	$A_4^1 B_4^0$	$A_5^1 B_5^0$	$A_6^1 B_6^0$
XY = 11	$A_1^1 B_1^1$	$A_2^1 B_2^1$	$A_3^1 B_3^1$	$A_4^1 B_4^1$	$A_5^1 B_5^1$	$A_6^1 B_6^1$

Entropy Bound - Private Randomness by Both Parties

- ▶ Proved that $|\mathcal{M}_{12}| \ge 6$ for non-interactive protocols with private randomness
- ▶ Next step : Prove $H(M_{12}) \ge \log_2 6$ for the same class of protocols
- ▶ Let $x \in \mathcal{X}$ and $S_x \subseteq \mathcal{Y}$ such that f(x,y) = f(x,y') for all $y, y' \in S_x$
- ▶ Without private randomness, it is known that (with x = 0)

$$H(M_{12}) \ge H(M_{13}|X=x) + \log_2 |S_x| = H(M_{13}) + 1 \ge \log_2 3 + 1$$

Proved that with private randomness,

$$H(M_{12}) \ge I(M_{12}; M_{13}|X=x) + \log_2 |S_x|$$

▶ Let $x \in \mathcal{X}$, $S_x = \{y_1, \dots y_{|S_x|}\}$, and $m_{12} \in \mathcal{M}_{12}$ such that $[a, b_i] \in \mathsf{supp}((M_{13}, M_{23})|m_{12}, x, y_i)$

▶ From the secrecy condition,

$$Pr(M_{13} = a, M_{23} = b_i|x, y_i) = Pr(M_{13} = a, M_{23} = b_i|x, y_1)$$

First show that

$$Pr(M_{13}=a|x) \geq |S_x| Pr(M_{12}=m_{12}) Pr(M_{13}=a|m_{12},x)$$

and hence whenever $a \in \text{supp}(M_{13}|M_{12}=m_{12},X=x)$,

$$p(m_{12}) \leq \frac{p(a|x)}{|S_x| \ p(a|m_{12},x)}$$

Denote supp $(M_{23}|m_{12}, y_i)$ as B_i

$$Pr(M_{13} = a|x) = Pr(M_{13} = a|x, y_1)$$

$$\geq \sum_{i=1}^{|S_x|} \sum_{b \in B_i} Pr(M_{13} = a, M_{23} = b|x, y_1)$$

$$= \sum_{i=1}^{|S_x|} \sum_{b \in B_i} Pr(M_{13} = a, M_{23} = b|x, y_i)$$

$$= \sum_{i=1}^{|S_x|} \sum_{b \in B_i} \sum_{m \in \mathcal{M}_{12}} Pr(M_{12} = m) Pr(M_{13} = a|m, x) Pr(M_{23} = b|m, y_i)$$

$$\geq \sum_{i=1}^{|S_x|} \sum_{b \in B_i} Pr(M_{12} = m_{12}) Pr(M_{13} = a|m_{12}, x) Pr(M_{23} = b|m_{12}, y_i)$$

$$= |S_x| Pr(M_{12} = m_{12}) Pr(M_{13} = a|m_{12}, x)$$

Denote supp $(M_{13}|M_{12}=m,X=x)$ as A(m,x) and $supp(M_{13}|X=x)$ as A(x)

$$H(M_{12}) = \sum_{m \in \mathcal{M}_{12}} p(m) \log_2 \left(\frac{1}{p(m)}\right)$$

$$= \sum_{m \in \mathcal{M}_{12}} \sum_{a \in A(m,x)} p(a|m,x)p(m) \log_2 \left(\frac{1}{p(m)}\right)$$

$$\geq \sum_{m \in \mathcal{M}_{12}} \sum_{a \in A(m,x)} p(a|m,x) p(m) \log_2 \left(\frac{|S_x| p(a|m,x)}{p(a|x)} \right)$$

$$= \sum_{a \in A(x)} \sum_{m \in \mathcal{M}_{12}} p(a|m, x) p(m) \log_2 \left(\frac{|S_x|}{p(a|x)}\right) - p(a|m, x) p(m) \log_2 \left(\frac{1}{p(a|m, x)}\right)$$

$$= \log_2 |S_x| + \sum_{a \in A(x)} p(a|x) \log_2 \left(\frac{1}{p(a|x)}\right) - \sum_{m \in \mathcal{M}_{12}} p(m) H(M_{13}|m,x)$$

$$= \log_2 |S_x| + H(M_{13}|X = x) - H(M_{13}|M_{12}, X = x)$$

$$= \log_2 |S_x| + H(M_{13}|X = x) - H(M_{13}|M_{12}, X = x)$$

= $I(M_{12}; M_{13}|X = x) + \log_2 |S_x|$

► Evaluating the bound

$$H(M_{12}) \ge I(M_{12}; M_{13}|X=x) + \log_2 |S_x|$$

For the protocol with randomness

$$H(M_{12}) \ge I(M_{12}; M_{13}|X = x) + \log_2 |S_x| = \log_2 3 + \log_2 3 = \log_2 9$$

- ▶ The protocol uses $H(M_{12}) = \log_2 12$
- ▶ For AND, choosing x = 0 and $S_x = \{0, 1\}$

$$H(M_{12}) \geq I(M_{12}; M_{13}|X=x) + 1$$

▶ It is not clear how to evaluate $I(M_{12}; M_{13}|X = x)$ in terms of input distributions

Relation with Distribution Design

- Problem motivated by secure multiparty computation, scret sharing
- ▶ Design a joint distribution on a set of random variables X_1, \ldots, X_n , that satisfies a set of constraints.
- ► Two types of constraints on sets of random variables $(X_{i_1}, \ldots, X_{i_d})$ and $(X_{i'_1}, \ldots, X_{i'_d})$
 - They have identical joint distributions

$$(X_{i_1},\ldots,X_{i_d})\equiv (X_{i'_1},\ldots,X_{i'_d})$$

They have disjoint support sets

$$(X_{i_1},\ldots,X_{i_d}) \parallel (X_{i'_1},\ldots,X_{i'_d})$$

Distribution Design - Secure Computation of AND

Random Variables:

- ▶ M_{13}^0, M_{13}^1 : Messages sent by Alice for inputs 0 and 1
- ▶ M_{23}^0, M_{23}^1 : Messages sent by Bob for inputs 0 and 1

Constraints:

$$(M_{13}^0, M_{23}^0) \equiv (M_{13}^0, M_{23}^1) \equiv (M_{13}^1, M_{23}^0)$$

$$(M_{13}^1, M_{23}^1) \parallel (M_{13}^0, M_{23}^0)$$

$$(M_{13}^1, M_{23}^1) \parallel (M_{13}^0, M_{23}^1)$$

$$(M_{13}^1, M_{23}^1) \parallel (M_{13}^1, M_{23}^0)$$

Use Lemma : $A_0 \subseteq [n], 0 < |A_0| = d < n$. Consider the constraints $\{A \mid\mid A_0: , A \subseteq [n], |A| = d, A \neq A_0\} \cup \{A \equiv A': A, A' \subseteq [n], |A| = |A'| = d, A, A' \neq A_0\}$. Then there exists a distribution design with $\lceil \log_2 |\operatorname{supp}(X_i)| \rceil$ at most $\min\{2d \cdot \log_2 n, n-1\}$

Distribution Design - Result and Limitations

Result:

- ▶ Distribution design exists for secure AND with $|\sup(M_{13})| = |\sup(M_{13}^0)| = |\sup(M_{13}^1)|$ at most 8
- ▶ For AND of k parties, $|\text{supp}(M_i)|$ at most 2^{2k-1}

Limitations:

- ▶ Known protocol requires only $|\text{supp}(M_i)| \sim O(k)$
- Used lemma includes unnecessary constraints of the form

$$(\textit{M}^0_{13}, \textit{M}^1_{13}) \, \| \, (\textit{M}^1_{13}, \textit{M}^1_{23})$$
 and $(\textit{M}^0_{13}, \textit{M}^1_{13}) \equiv (\textit{M}^0_{13}, \textit{M}^0_{23})$

- ▶ Proof of lemma uses construction that may not be optimal in general
- Does not explicitly use the common randomness variables

Distribution Design - Secure Multiparty Computation

- Formulate secure multiparty computation of any function $f: \mathcal{X}_1 \times \ldots \times \mathcal{X}_k \to \{0,1\}$ as a distribution design
- ► Optimal bounds for distribution design will also solve secure multiparty computation
- ▶ Choose the random variables $\{M_i^x : x \in \mathcal{X}_i, i = 1, ..., n\}$, and the constraints as

$$(M_1^{x_1}, \dots, M_k^{x_k}) \equiv (M_1^{x'_1}, \dots, M_k^{x'_k}) \quad \text{if } f(x_1, \dots, x_k) = f(x'_1, \dots, x'_k)$$
$$(M_1^{x_1}, \dots, M_k^{x_k}) \parallel (M_1^{x'_1}, \dots, M_k^{x'_k}) \quad \text{if } f(x_1, \dots, x_k) \neq f(x'_1, \dots, x'_k)$$

Future Work

- Generic bounds that can extend to functions other than AND as well
- Formalize the counting argument proof in a systematic way
- ▶ Entropy bounds Evaluating $I(M_{12}; M_{13}|X = x)$ for general functions
- Extending entropy bounds to multiparty secure computation
- Analyzing this problem as a distribution design problem, with added constraints for using the common randomness variables.
- Alternately, consider a support design problem

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Questions?

Thank You!