# Strong XOR Lemma for Communication with Bounded Rounds

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#### For function $f : \mathbb{Z} \to \{0, 1\}$ , its *n*-fold XOR $f^{\oplus n} : \mathbb{Z}^n \to \{0, 1\}$ is:

$$f^{\oplus n}(Z_1,\ldots,Z_n)=f(Z_1)\oplus\cdots\oplus f(Z_n)$$

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This talk: "CC of f" vs "CC of  $f^{\oplus n}$ "

Suppose f can be computed using resource C w.p. 2/3 Compute n copies independently and output their XOR Suppose f can be computed using resource C w.p. 2/3 Compute n copies independently and output their XOR Use  $n \cdot C$  resource in total, and succeed w.p.  $1/2 + \exp(-\Theta(n))$  Suppose f can be computed using resource C w.p. 2/3 Compute n copies independently and output their XOR Use  $n \cdot C$  resource in total, and succeed w.p.  $1/2 + \exp(-\Theta(n))$ If this is the best possible, then

• moderately hard Boolean-valued  $f \implies$  very hard Boolean-valued  $f^{\oplus n}$ 

A strong XOR lemma (for a model of computation and a class of functions): " $f^{\oplus n}$  cannot be computed much better than solving all instances independently" A strong XOR lemma (for a model of computation and a class of functions): " $f^{\oplus n}$  cannot be computed much better than solving all instances independently"

Previous XOR lemmas:

- query complexity [Dru'12, BKLS'20]
- w/o n times more resource: circuit complexity [Yao'82], streaming alg [AN'21]
- w/o exponentially small adv: information complexity [BBCR'10]

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- communication complexity & functions with small discrepancy [Shaltiel'03]
- . . .

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*n*-fold XOR function:

$$f^{\oplus n}(X_1,\ldots,X_n,Y_1,\ldots,Y_n) = f(X_1,Y_1)\oplus\cdots\oplus f(X_n,Y_n)$$



Let  $\mathbf{R}_q^{(r)}(f)$  be the min communication cost to compute f in r rounds with prob q.

Theorem

For any f and r, we must have

$$\mathbf{R}_{1/2+2^{-n}}^{(r)}(f^{\oplus n}) \ge n \cdot \left(r^{-O(r)} \cdot \mathbf{R}_{2/3}^{(r)}(f) - 1\right).$$

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#### Remarks:

• for constant r: 
$$\mathbf{R}_{1/2+2^{-n}}^{(r)}(f^{\oplus n}) \ge \Omega(n \cdot (\mathbf{R}_{2/3}^{(r)}(f) - O(1)))$$

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- "-O(1)" is needed:  $f(X_i, Y_i) = X_{i,1} \oplus Y_{i,1}$  (XOR of 1st bit)

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[BBCR'10]: XOR lemma for info complexity  $\iff$  starting point of our proof (with const adv instead of  $2^{-n}$ )

[BR'11]: for const r, information  $\implies$  communication Imply: for const r,  $\mathbf{R}_{2/3}^{(r)}(f^{\oplus n}) \ge \Omega(n \cdot (\mathbf{R}_{2/3}^{(r)}(f) - O(1)))$ 

### Distributional strong XOR lemma

We also prove a strong XOR lemma w.r.t. a fixed input distribution  $\mu$ :

#### Theorem

If every r-round C-bit comm. protocol computes f under input dist.  $\mu$  w.p. at most

 $1/2 + \alpha/2,$ 

then every r-round  $o(r^{-1}nC)$ -bit protocol computes  $f^{\oplus n}$  under  $\mu^n$  w.p. at most

 $1/2 + \alpha^{\Omega(n)}/2,$ 

where  $\alpha < r^{-\omega(r)}$  and  $C > \omega(\log(1/\alpha))$ .

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distributional strong XOR lemma + Yao's minimax + repetition  $\implies$  main theorem

Rest of the talk, focus on distributional strong XOR lemma:

- alternative view of the XOR lemma for information complexity [BBCR'10]
- obtaining exponentially small advantage

## Information complexity

input distribution  $\mu$  + protocol defines a joint distribution  $\pi$  over  $(X, Y, R, \mathbf{M})$ ...



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information complexity of f under  $\mu$ : min information cost to compute f

[BBCR'10]: <u>if</u> info complexity of  $f^{\oplus n}$  under  $\mu^n$  is  $\leq I$ , <u>then</u> info complexity of f under  $\mu$  is  $\leq I/n + O(1)$ (assuming success probability 1 for now)





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[BBCR'10]: <u>if</u> info complexity of  $f^{\oplus n}$  under  $\mu^n$  is  $\leq I$ , <u>then</u> info complexity of f under  $\mu$  is  $\leq I/n + O(1)$ 

an alternative view of their proof:

- fix  $\pi$  for  $f^{\oplus n}$  with info cost I
- "decompose"  $\pi$  into  $\pi_n$  for f and info cost  $l_1$  and  $\pi_{< n}$  for  $f^{\oplus n-1}$  with info cost  $l_2$ such that  $l_1 + l_2 = l + O(1)$





 $\frac{\text{Input: 1 pair}}{\text{Protocol } \pi_n}$ 

- view input as  $X_n$  and  $Y_n$
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<u>Cost</u>:  $I(X_n; \mathbf{M} \mid Y, R) + 1$  (1st term)



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Cost:  $I(X_n; \mathbf{M} \mid Y, R) + 1$  (1st term)



 $\frac{\text{Input: } n-1 \text{ pairs}}{\text{Protocol } \pi_{< n}}$ 

- view input as  $X_{< n}$  and  $Y_{< n}$
- publicly sample  $X_n$
- Bob privately samples  $Y_n$  cond. on  $X_n$
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# **Decomposition of** $\pi$



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Cost:  $I(X_{e}; \mathbf{M} \mid Y, R) + 1$  (1st term) Cost:  $I(X_{e}; \mathbf{M} \mid X_{n}, Y, R)$  (1st term)

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1st terms in costs sum up to  $I(X; \mathbf{M} \mid Y, R) + 1$  by chain rule

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- run  $\pi$  and Bob sends  $f(X_n, Y_n)$

1st terms in costs sum up to  $I(X; \mathbf{M} \mid Y, R) + 1$  by chain rule 2nd term is similar; info costs of  $\pi_{< n}$  and  $\pi_n$  sum up to I + O(1)



iteratively decomposing  $\pi_{< n}$  gives n protocols for f

• *i*-th last: the original protocol when it embeds input into  $(X_i, Y_i)$ 

for the <u>same</u> underlying distribution of  $(X, Y, R, \mathbf{M})$ , we view different parts of it as inputs, public randomness, transcript (private randomness not important)

- $\pi$ : inputs (X, Y), public randomness R, transcript M
- $\pi_n$ : inputs  $(X_n, Y_n)$ , public rand.  $(R, Y_{< n})$ , transcript  $(\mathbf{M}, f^{\oplus n-1}(X_{< n}, Y_{< n}))$
- $\pi_{<n}$ : inputs  $(X_{<n}, Y_{<n})$ , public randomness  $(R, X_n)$ , transcript  $(M, f(X_n, Y_n))$

given a protocol computing  $f^{\oplus n}$  w.p. 2/3 under  $\mu^n$  with cost o(nC)then there is a protocol computing f w.p. 2/3 under  $\mu$  with cost  $\leq C$  To prove strong XOR lemma, need to show: <u>given</u> a protocol computing  $f^{\oplus n}$  w.p.  $1/2 + \alpha^{o(n)}/2$  under  $\mu^n$  with cost o(nC)then there is a protocol computing f w.p.  $1/2 + \alpha/2$  under  $\mu$  with cost  $\leq C$  To prove strong XOR lemma, need to show: <u>given</u> a protocol computing  $f^{\oplus n}$  w.p.  $1/2 + \alpha^{o(n)}/2$  under  $\mu^n$  with cost o(nC)<u>then</u> there is a protocol computing f w.p.  $1/2 + \alpha/2$  under  $\mu$  with cost  $\leq C$ 

Main challenge: design a decomposition that increases the advantage

- given **W**, one can predict f w.p.  $1/2 + adv(f | \mathbf{W})/2$
- $\operatorname{adv}(b_1 \oplus b_2) = \operatorname{adv}(b_1) \cdot \operatorname{adv}(b_2)$

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End of  $\pi_n$ , Alice knows  $(X_n, Y_{< n}, R, \mathbf{M})$ 



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End of  $\pi_{< n}$ , Bob knows  $(X_n, Y_{< n}, R, \mathbf{M})$ 



- given **W**, one can predict f w.p.  $1/2 + adv(f | \mathbf{W})/2$
- $\operatorname{adv}(b_1 \oplus b_2) = \operatorname{adv}(b_1) \cdot \operatorname{adv}(b_2)$



Key obs:  $f(X_n, Y_n)$  and  $f^{\oplus n-1}(X_{\leq n}, Y_{\leq n})$  are independent cond. on  $(X_n, Y_{\leq n}, R, M)$ 

### Benefit of the alternative view



Key obs:  $f(X_n, Y_n)$  and  $f^{\oplus n-1}(X_{\leq n}, Y_{\leq n})$  are independent cond. on  $(X_n, Y_{\leq n}, R, M)$ 

### Benefit of the alternative view



Key obs:  $f(X_n, Y_n)$  and  $f^{\oplus n-1}(X_{< n}, Y_{< n})$  are independent cond. on  $(X_n, Y_{< n}, R, \mathbf{M})$ Since  $f^{\oplus n}(X, Y) = f^{\oplus n-1}(X_{< n}, Y_{< n}) \oplus f(X_n, Y_n)$ ,

 $\operatorname{adv}(f(X_n, Y_n) \mid X_n, Y_{< n}, R, \mathsf{M}) \cdot \operatorname{adv}(f^{\oplus n-1}(X_{< n}, Y_{< n}) \mid X_n, Y_{< n}, R, \mathsf{M})$  $= \operatorname{adv}(f^{\oplus n}(X, Y) \mid X_n, Y_{< n}, R, \mathsf{M})$ 

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Relate adv of  $\pi_n$  and adv of  $\pi_{< n}$  to adv of  $\pi$ 

 $\operatorname{adv}(f(X_n, Y_n) \mid X_n, Y_{< n}, R, \mathsf{M}) \cdot \operatorname{adv}(f^{\oplus n-1}(X_{< n}, Y_{< n}) \mid X_n, Y_{< n}, R, \mathsf{M})$  $= \operatorname{adv}(f^{\oplus n}(X, Y) \mid X_n, Y_{< n}, R, \mathsf{M})$ 

Relate adv of  $\pi_n$  and adv of  $\pi_{< n}$  to adv of  $\pi$ 

If  $\pi_n$  does not have "high success prob", then adv of  $\pi_{< n}$  is larger than adv of  $\pi$  by a factor

 $\operatorname{adv}(f(X_n, Y_n) \mid X_n, Y_{< n}, R, \mathsf{M}) \cdot \operatorname{adv}(f^{\oplus n-1}(X_{< n}, Y_{< n}) \mid X_n, Y_{< n}, R, \mathsf{M})$  $= \operatorname{adv}(f^{\oplus n}(X, Y) \mid X_n, Y_{< n}, R, \mathsf{M})$ 

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If  $\pi_n$  does not have "high success prob", then adv of  $\pi_{< n}$  is larger than adv of  $\pi$  by a factor

• decomposition increases the advantage

### Proof strategy:

- 1. given  $\pi$  for  $f^{\oplus n}$ , decompose into  $\pi_n$  for f and  $\pi_{< n}$  for  $f^{\oplus n-1}$
- 2. prove:

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3. if  $\pi$  has "low cost" and non-trivial adv: iterative decomposition gives a good protocol for f

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Thank you for listening!