

# Small Error Versus Unbounded Error Protocols in the NOF Model

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# NOF Model

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- Communication by writing on blackboard (broadcast).



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$$PP_k(f) \equiv \min_{\epsilon} \left[ R_{\epsilon}^{pub}(f) + \log \left( \frac{1}{\epsilon} \right) \right], \quad UPP_k(f) \equiv \min_{\epsilon} [R_{\epsilon}^{priv}(f)]$$

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Define  $(U)PP_k^{cc} = \{f : (U)PP_k(f) = polylog(n)\}$

- Not hard:  $PP_k^{cc} \subseteq UPP_k^{cc}$ . (Follows from Newman's Theorem).

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- $PP_k^{cc} \subsetneq UPP_k^{cc}$ ,  $k \leq O(\log \log(n))$  (follows from Beigel ['94] + Sherstov ['14]). Shows an  $\Omega(n^{1/3})$  lower bound.



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## Main results

- $PP_k^{cc} \subsetneq UPP_k^{cc}, k \leq \Theta(\log(n))$ .
- $\Omega\left(\frac{\sqrt{n}}{4^k}\right)$   $PP_k$  lower bound for functions in  $UPP_k^{cc}$ .
- There exists a function that is computed very efficiently by  $THR \circ PAR_{k+1}$  circuits but requires  $2^{\Omega\left(\frac{\sqrt{n}}{4^k}\right)}$  size to be computed by depth-three circuits of the form  $MAJ \circ THR \circ ANY_k$ .

# Target function

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Definition (Goldmann, Håstad, Razborov ['92])

Let

$$P(x, y_1, \dots, y_k) \equiv \sum_{i=0}^{n-1} \sum_{j=0}^{n4^k-1} 2^i y_{1j} \dots y_{kj} (x_{i,2j} + x_{i,2j+1})$$

where  $x \in \{\pm 1\}^{2n^2 4^k}$ ,  $y_i \in \{\pm 1\}^{n4^k}$  for each  $i$ .

Then,  $\text{GHR}_k^N(x, y_1, \dots, y_k) \equiv \text{sgn}(2P(x, y_1, \dots, y_k) + 1)$ , where  $N = 2n^2 4^k$ .

# Discrepancy and Cylinder Intersections

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## Lemma (Folklore)

$$R_\epsilon^{\text{pub}}(f) \geq \log(2\epsilon / \min_\mu \text{Disc}_\mu^k(f)).$$

Thus,  $\text{PP}_k$  lower bounds exactly correspond to discrepancy upper bounds.

# Discrepancy

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Let  $f : X_1 \times \cdots \times X_k \rightarrow \{-1, 1\}$ .

## Definition

Let  $\mu$  be a distribution on  $X_1 \times \cdots \times X_k$ . The discrepancy of  $f$  according to  $\mu$ ,  $Disc_\mu^k(f)$  is

$$\max_S |\mu(f^{-1}(1) \cap S) - \mu(f^{-1}(-1) \cap S)|$$

where the maximum is taken over all cylinder intersections  $S$ .

# Discrepancy under product distributions

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## Lemma (Folklore)

Let  $f : X \times Y_1 \times \dots \times Y_k \rightarrow \{-1, 1\}$ , and  $\mu$  any product distribution. Then,

$$(\text{Disc}_\mu^{k+1}(f))^{2^k} \leq \mathbb{E}_{y_1^0, y_1^1, \dots, y_k^0, y_k^1} \left[ \mathbb{E}_X \prod_{a \in \{0,1\}^k} f(x, y_1^{a_1}, \dots, y_k^{a_k}) \right]$$

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- $x$ 's are distributed such that  $A_j = \frac{1}{2} \sum_{i=0}^{n-1} 2^i (x_{i,2j} + x_{i,2j+1})$  is binomially distributed for each  $0 \leq j \leq n4^k - 1$ .

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- Note  $\text{GHR}_k^N(x, y_1, \dots, y_k) = \text{sgn}(\sum_{j=0}^{n4^k} A_j y_{1j} \dots y_{kj})$ .

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- Will show that for many fixings of the  $y_i^j$ 's, the inner expectation is small.
- Analyze the number and size of integral solutions to Hadamard constraints.
- Use anticoncentration properties of binomial distribution.

# Circuit Lower Bounds



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*For  $f \in \text{MAJ} \circ \text{SYM} \circ \text{ANY}_k$  of size  $s$ , and any partition of the input bits amongst  $k + 1$  players, there exists a randomized protocol computing  $f$  with advantage  $\Omega(1/s)$  and cost  $O(k \log(s))$ .*

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### Lemma (GHR[92])

$$\text{MAJ} \circ \text{THR} \subseteq \text{MAJ} \circ \text{MAJ}$$

- $\text{GHR}_k$  requires  $2^{\Omega\left(\frac{\sqrt{n}}{4^k}\right)}$  size to be computed by  $\text{MAJ} \circ \text{THR} \circ \text{ANY}_k$  circuits.

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- $\Omega(n)$   $PP_k$  lower bound for an explicit function in  $UPP_k$  (open for 2 player case too)?
- Is  $GHR_k$  hard for  $k > \log(n)$  players?
- Can we find an explicit  $f$  not in  $UPP_3^{cc}$ ? This will show that  $f$  is not in polynomial size  $THR \circ THR$  (Hansen, Podolskii ['15]).

# Thank You!