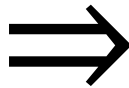


Lower Bounds for Testing Properties of Functions on Hypergrids

Grigory Yaroslavtsev

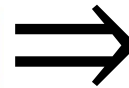
<http://grigory.us>



BROWN



ICERM



Penn
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Joint with:

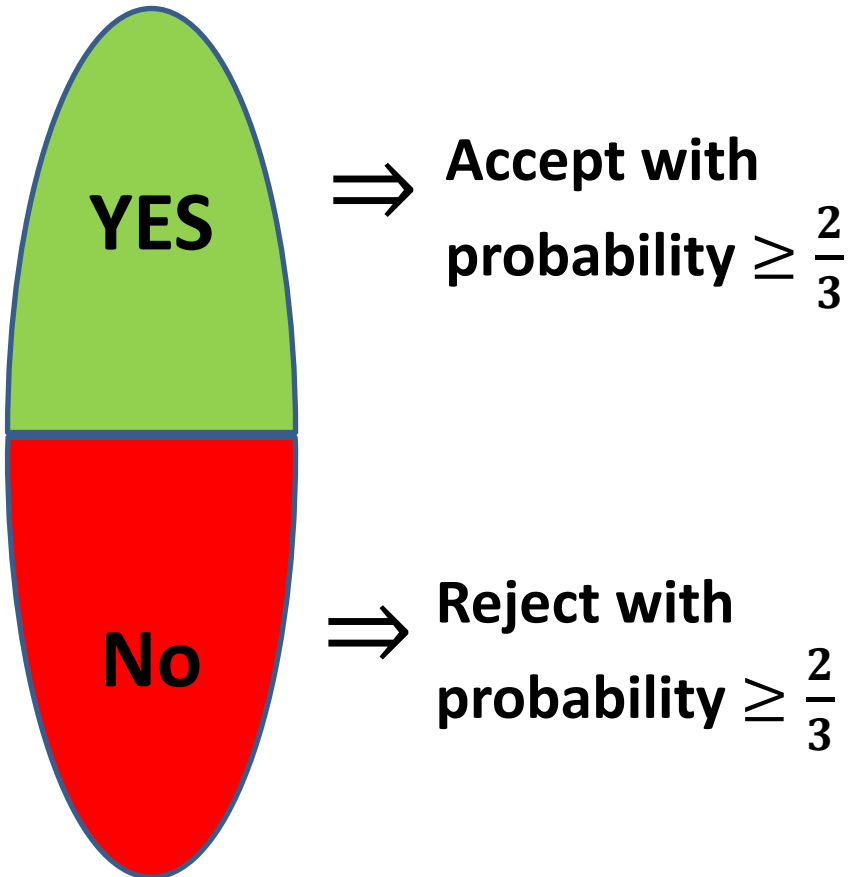
Eric Blais (MIT)

Sofya Raskhodnikova (PSU)

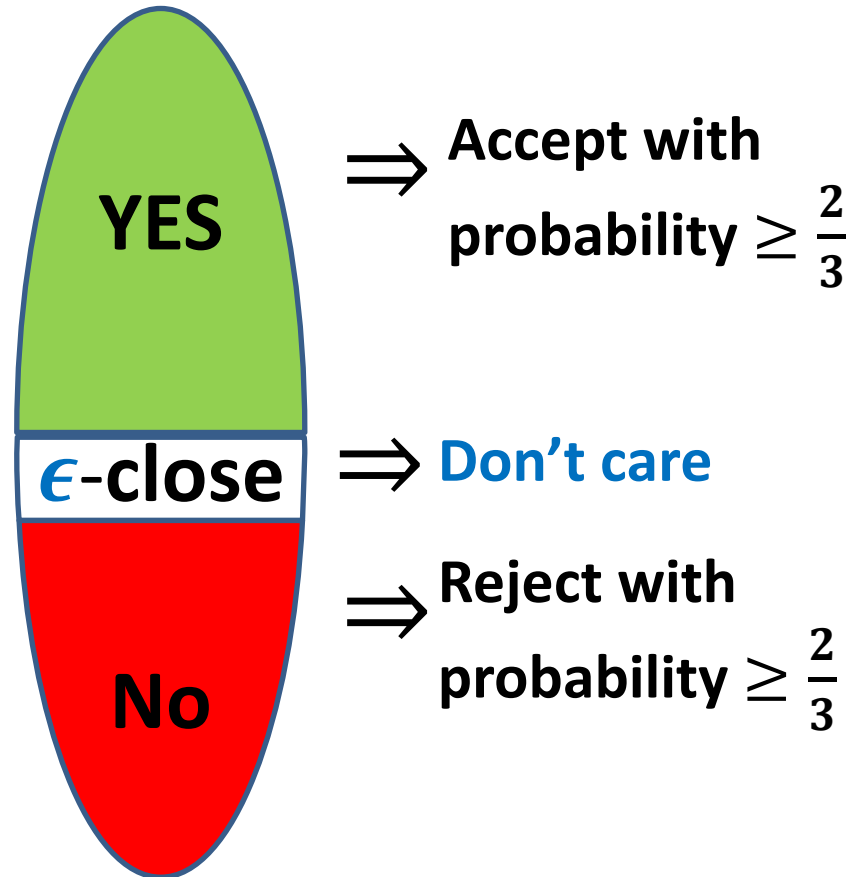
Property Testing

[Goldreich, Goldwasser, Ron, Rubinfeld, Sudan]

Randomized algorithm



Property tester



ε-close : $\leq \epsilon$ fraction can be changed to become YES

Ultra-fast Approximate Decision Making



Property Testing

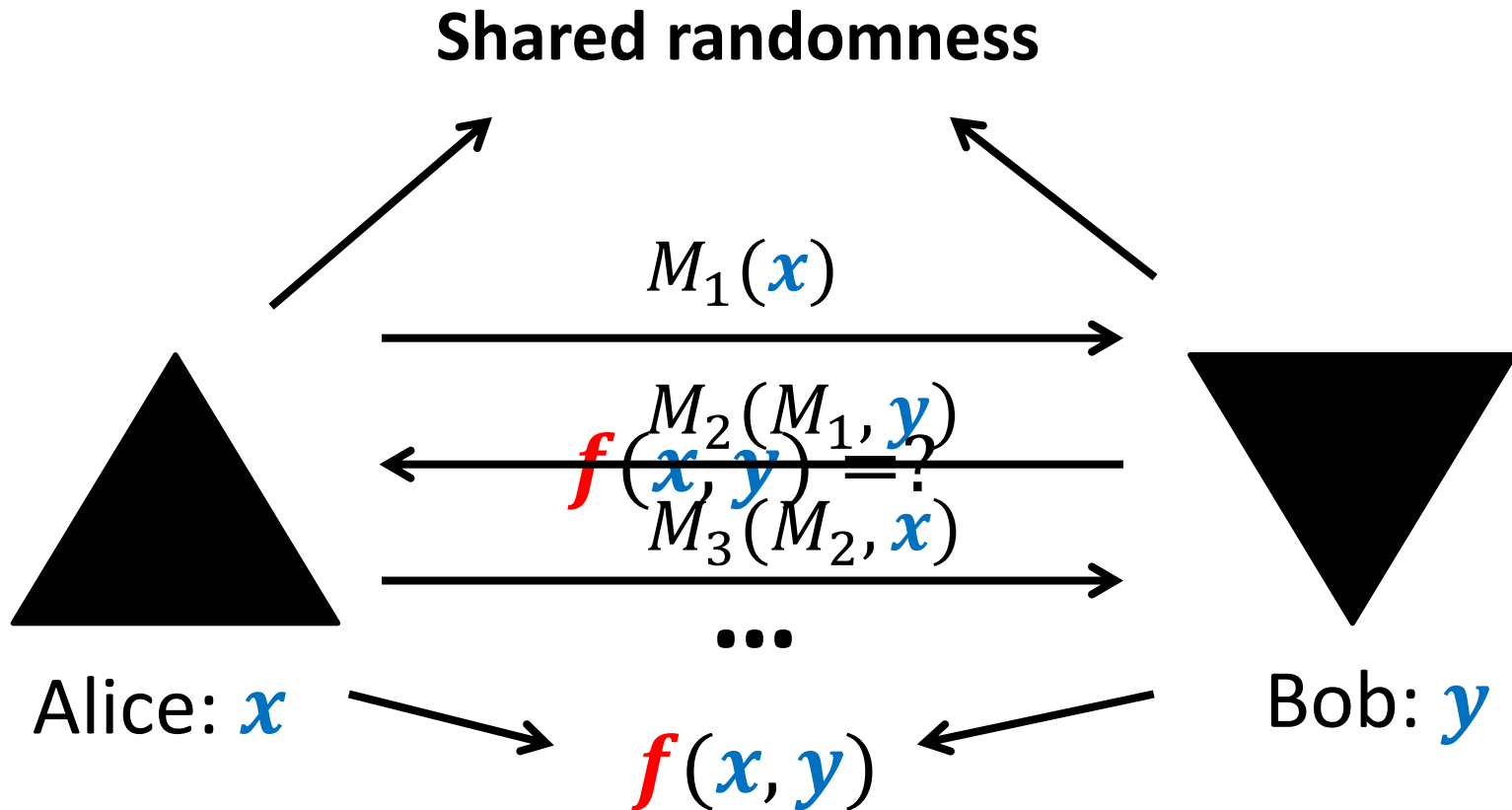
[Goldreich, Goldwasser, Ron, Rubinfeld, Sudan]

Property P = set of **YES** instances

Query complexity of testing P :

- $Q_{\epsilon}(P)$ = Adaptive queries
- $Q_{\epsilon}^{na}(P)$ = Non-adaptive (all queries at once)
- $Q_{\epsilon}^r(P)$ = Queries in r rounds ($Q_{\epsilon}^{na}(P) = Q_{\epsilon}^1(P)$)

Communication Complexity [Yao'79]



- $R(f)$ = min. communication (error 1/3)
- $R^k(f)$ = min. k -round communication (error 1/3)

$k/2$ -disjointness \Rightarrow k -linearity

[Blais, Brody, Matulef'11]

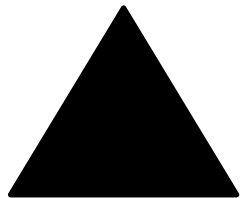
- k -linear function: $\{0,1\}^n \rightarrow \{0,1\}$

$$\bigoplus_{i \in S} x_i = x_{i_1} \oplus x_{i_2} \oplus \cdots \oplus x_{i_k}$$

where $|S| = k$

- $k/2$ -Disjointness: $S, T \subseteq [n]$, $|S| = |T| = \frac{k}{2}$

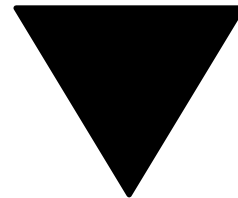
$$f(S, T) = 1, \text{ iff } |S \cap T| = 0.$$



Alice:

$$S \subseteq [n], |S| = k/2$$

$$f: |S \cap T| = 0?$$

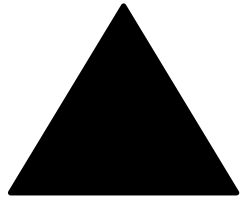


Bob:

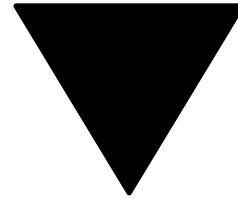
$$T \subseteq [n], |T| = k/2$$

$k/2$ -disjointness \Rightarrow k -linearity

[Blais, Brody, Matulef'11]



$$\chi = \chi_S \oplus \chi_T$$



$$S \subseteq [n], |S| = k/2$$

$$\chi_S = \bigoplus_{i \in S} x_i$$

$$T \subseteq [n], |T| = k/2$$

$$\chi_T = \bigoplus_{i \in T} x_i$$

- $S \cap T = \emptyset \Rightarrow \chi$ is k -linear
- $S \cap T \neq \emptyset \Rightarrow \chi$ is $(< k)$ -linear, $1/2$ -far from k -linear
- Test χ for k -linearity using shared randomness
- To evaluate $\chi(x)$ exchange $\chi_S(x)$ and $\chi_T(x)$ (2 bits)
- $\mathbf{R}\left(\frac{k}{2}\text{-Disjointness}\right) \leq 2 \cdot \mathbf{Q}_{1/2}(k\text{-Linearity})$

k -Disjointness

- $R(k\text{-Disjointness}) = \Theta(k)$ [Razborov, Hastad-Wigderson]
- $R^1(k\text{-Disjointness}) = \Theta(k \log k)$
[Folklore + Dasgupta, Kumar, Sivakumar'12; Buhrman, Garcia-Soriano, Matsliah, De Wolf'12]
- $R^r(k\text{-Disjointness}) = \Theta(k \text{ilog}^r k)$,
where $\text{ilog}^r k = \underbrace{\log \log \dots \log k}_{r \text{ times}}$ [Saglam, Tardos'13]

$$\Omega(k \text{ilog}^r k) = Q_{1/2}^r(k\text{-Linearity})$$

- $R(k\text{-Disjointness}) = \alpha k + o(k)$ [Braverman, Garg, Pankratov, Weinstein'13]

Property testing lower bounds via CC

- Monotonicity, Juntas, Low Fourier degree, Small Decision Trees [Blais, Brody, Matulef'11]
- Small-width OBDD properties [Brody, Matulef, Wu'11]
- Lipschitz property [Jha, Raskhodnikova'11]
- Codes [Goldreich'13, Gur, Rothblum'13]
- Number of relevant variables [Ron, Tsur'13]

(Almost) all: Boolean functions over Boolean hypercube

Functions $[m]^n \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

$M_{m,n}$ = monotone functions over $[m]^n$

$$Q^1(M_{m,n}) = \Omega(n \log m)$$

Previous for monotonicity on the line ($n = 1$):

- $Q^1(M_{m,1}) = \Theta(\log m)$ [Ergun, Kannan, Kumar, Rubinfeld, Viswanathan'00]
- $Q(M_{m,1}) = \Omega(\log m)$ [Fischer'04]

Functions $[m]^n \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

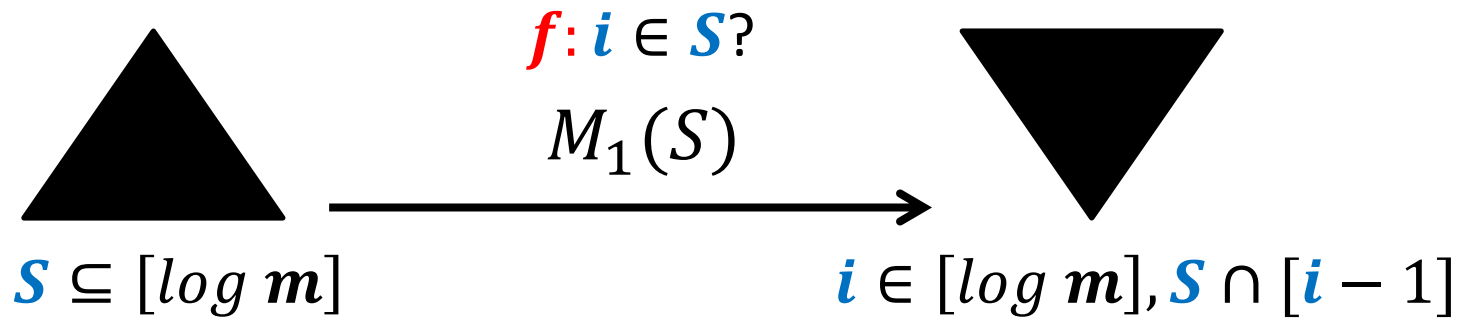
- **Proof ideas:**

- Reduction from Augmented Index (widely used in streaming, e.g [Jayram, Woodruff'11; Molinaro, Woodruff, Y.'13])
- Fourier analysis over $\{0,1\}^n$ basis of characters => Fourier analysis over $[m]^n$: basis of Walsh functions

- **Case $n = 1$:** Any non-adaptive tester for monotonicity of $f: [m] \rightarrow [r]$ has complexity $\Omega(\min(\log m, \log r))$

Functions $[m] \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

- **Augmented Index: $S; (i, S \cap [i - 1])$**



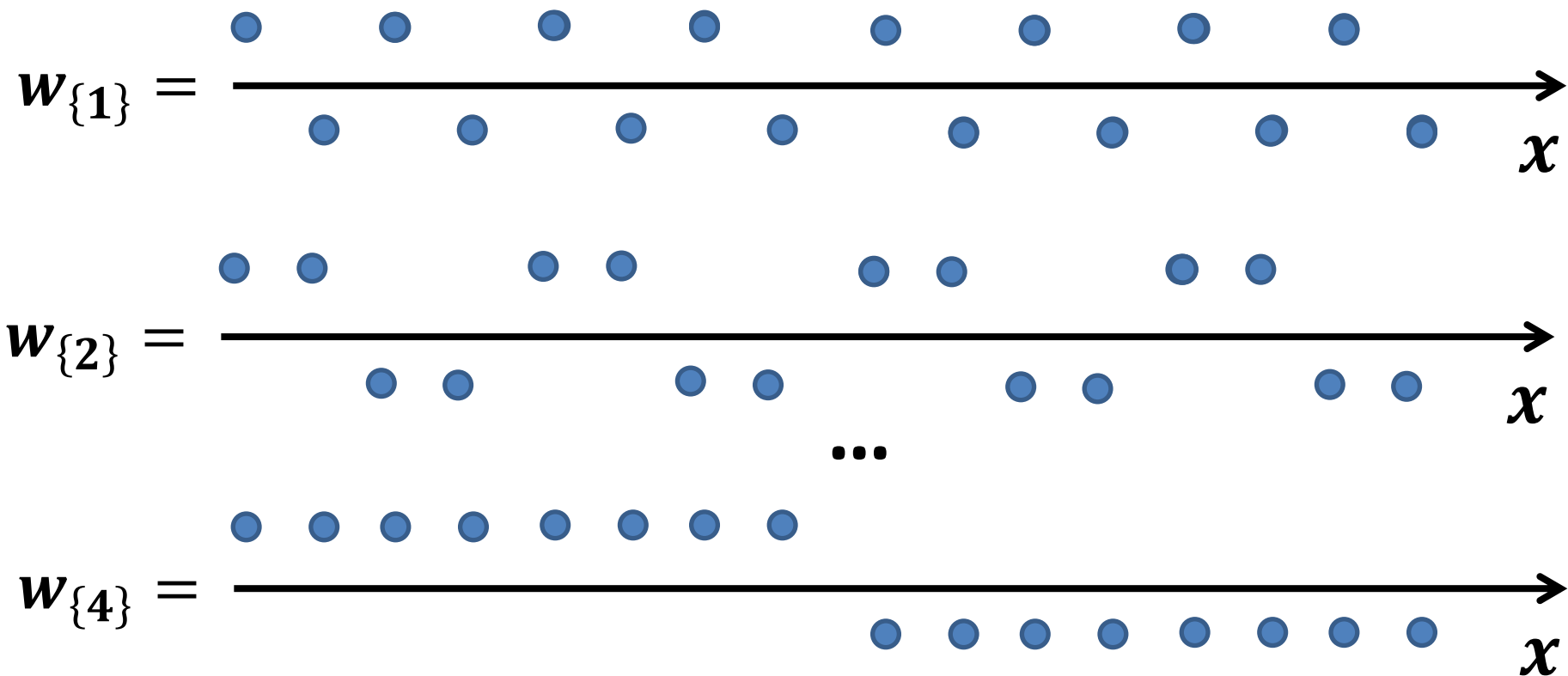
- $R^1[\text{Augmented Index}] = \Omega(|S|)$ [Miltersen, Nisan, Safra, Wigderson, 98]

Functions $[m] \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

Walsh functions: For $S \subseteq [\log m]$, $w_S: [m] \rightarrow \{-1, 1\}$:

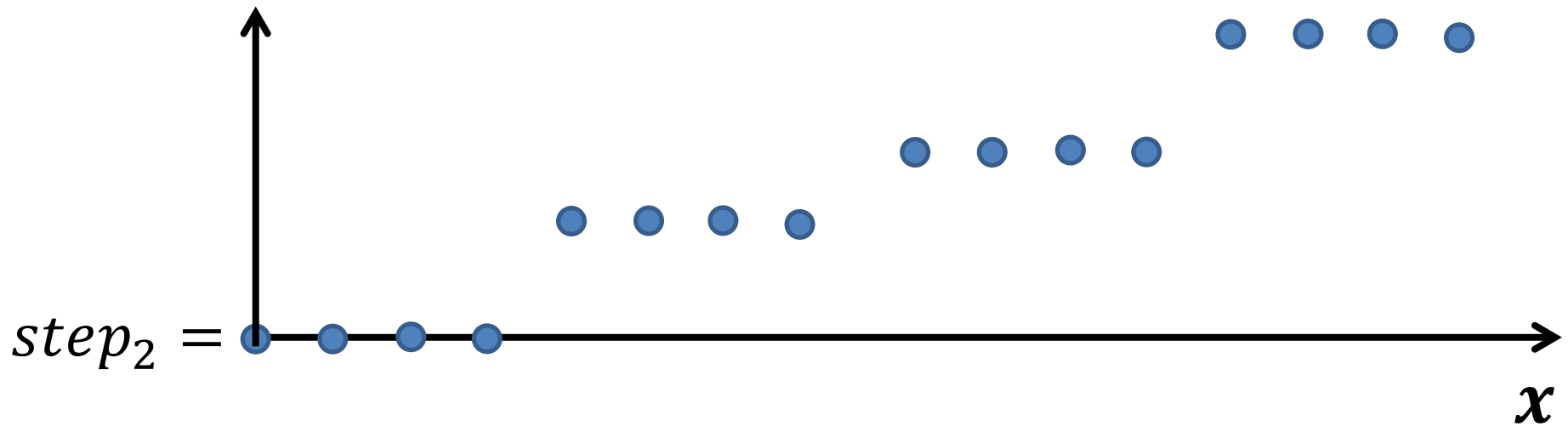
$$w_S(x) = \prod_{i \in S} (-1)^{x_i},$$

where x_i is the i -th bit of x .



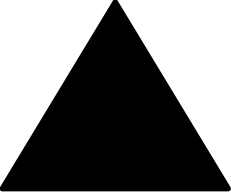
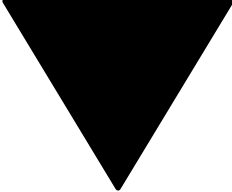
Functions $[m] \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

Step functions. For $i \in [\log m]$: $step_i: [m] \rightarrow \left[\frac{m}{2^i}\right]$:
 $step_i(x) = \lceil x/2^i \rceil$



Functions $[m] \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

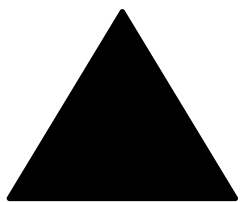
- **Augmented Index** \Rightarrow Monotonicity Testing


$$\begin{aligned} \chi &= w_{S \cap [i, \dots, \log m]} + 2 \text{step}_i \\ &= w_S \oplus w_{S \cap [i-1]} + 2 \text{step}_i \end{aligned}$$


$S \subseteq [\log m]$ $i \in [\log m], S \cap [i-1]$

- $i \notin S \Rightarrow \chi$ is monotone
- $i \in S \Rightarrow \chi$ is $\frac{1}{4}$ -far from monotone
- Only i -th frequency matters: higher frequencies are cancelled, lower don't affect monotonicity
- Thus, $Q^1(M_{m,1}) = \Omega(\log m)$

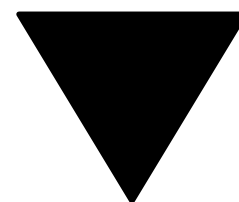
Functions $[m]^n \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]



$$S \subseteq [n \log m]$$



$$S_1, \dots, S_n \subseteq [\log m]$$



$$i \in [n \log m], S \cap [i - 1]$$



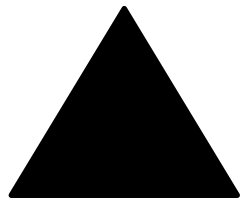
$$(m, m, \dots, m, i_{j^*}, 0, 0, \dots, 0)$$

$$S_1, \dots, S_{j^*-1}, S_{j^*} \cap [i_{j^*} - 1]$$

Embed into j^* -th coordinate using n -dimensional Walsh and step functions:

- Walsh functions: $w_S(x_1, \dots, x_n) = \prod_{j=1}^n w_{S_j}(x_j)$
- Step functions: $\text{step}_i(x_1, \dots, x_n) = \sum_{j=1}^n \text{step}_j(x_j)$

Functions $[m]^n \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]



$$\mathcal{S}_1, \dots, \mathcal{S}_n \subseteq [\log m]$$

$$\chi(x_1, \dots, x_n) =$$

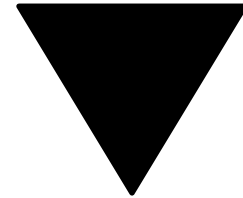
$$w_{\mathcal{S}_{j^*} \cap [i_{j^*}, \dots, \log m]}(x_{j^*}) \oplus \prod_{j=j^*+1}^n w_{\mathcal{S}_j}(x_j) + 2 \mathbf{step}_i(x_1, \dots, x_n) =$$

$$w_{\mathcal{S}}(x_1, \dots, x_n) \oplus \prod_{j=1}^{j^*-1} w_{\mathcal{S}_j}(x_j) +$$

$$2 \sum_{j=1}^{j^*-1} \mathbf{step}_m(x_j) + 2 \mathbf{step}_{i_{j^*}}(x_{j^*}) + 2 \sum_{j=j^*+1}^n \mathbf{step}_0(x_j)$$

- Walsh functions: $w_{\mathcal{S}}(x_1, \dots, x_n) = \prod_{j=1}^n w_{\mathcal{S}_j}(x_j)$

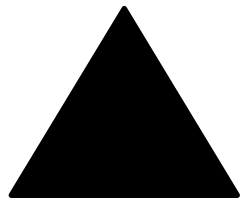
- Step functions: $\mathbf{step}_i(x_1, \dots, x_n) = \sum_{j=1}^n \mathbf{step}_j(x_j)$



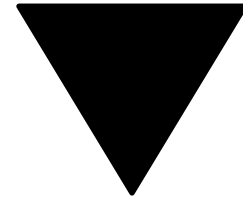
$$(m, m, \dots, m, i_{j^*}, 0, 0, \dots, 0)$$

$$\mathcal{S}_1, \dots, \mathcal{S}_{j^*-1}, \mathcal{S}_{j^*} \cap [i_{j^*} - 1]$$

Functions $[m]^n \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]



$$S_1, \dots, S_n \subseteq [\log m]$$



$$(m, m, \dots, m, i_{j^*}, 0, 0, \dots, 0)$$

$$S_1, \dots, S_{j^*-1}, S_{j^*} \cap [i_{j^*} - 1]$$

$$\chi(x_1, \dots, x_n) =$$

$$w_S(x_1, \dots, x_n) \oplus \prod_{j=1}^{j^*-1} w_{S_j}(x_j) + 2step_{i_{j^*}}(x_{j^*}) + 2 \sum_{j=j^*+1}^n x_j$$

- Only coordinate j^* matters:
 - Coordinates $< j^*$ cancelled by Bob's Walsh terms
 - Coordinates $> j^*$ cancelled by Bob's Step terms
 - Coordinate j^* behaves as in the $n = 1$ case

Functions $[m]^n \rightarrow \mathbb{R}$ [Blais, Raskhodnikova, Y.]

- $M_{m,n}$ = monotone functions over $[m]^n$
 $Q^1(M_{m,n}) = \Omega(n \log m)$
- $L_{m,n}$ = c -Lipschitz functions over $[m]^n$
- $C_{m,n}^S$ = separately convex functions over $[m]^n$
- $M_{m,n}^k$ = monotone axis-parallel k -th derivative over $[m]^n$
- $C_{m,n}$ = convex functions over $[m]^n$
 - Can't be expressed as a property of axis-parallel derivatives!

Thm. [BRY] For all these properties $Q^1 = \Omega(n \log m)$

These bounds are optimal for $M_{m,n}$ and $L_{m,n}$ [Chakrabarty, Seshadhri, '13]

Open Problems

- Adaptive bounds and round vs. query complexity tradeoffs for functions $[m]^n \rightarrow \mathbb{R}$
 - Only known: $Q(M_{m,n}) = \Omega(n \log m)$ [Fischer'04; Chakrabarty Seshadhri'13]
- Inspired by connections of CC and Information Complexity
 - Direct information-theoretic proofs?
 - Round vs. query complexity tradeoffs in property testing?
- Testing functions $[0, 1]^n \rightarrow \mathbb{R}$
 - L_p -testing model [Berman, Raskhodnikova, Y. '14]
 - Testing convexity: $2^{O(n \log n)}$ vs. $\Omega(n)$?