

Gap-Hamming-Distance: The Journey to an Optimal Lower Bound

Amit Chakrabarti

DARTMOUTH COLLEGE

Main result joint with

Oded Regev, TEL AVIV UNIVERSITY

Sublinear Algorithms Workshop at Bertinoro, May 2011

The Gap-Hamming-Distance Problem

Input: Alice gets $x \in \{0, 1\}^n$, Bob gets $y \in \{0, 1\}^n$.

Output:

- $\text{GHD}(x, y) = 1$ if $\Delta(x, y) > \frac{n}{2} + \sqrt{n}$
- $\text{GHD}(x, y) = 0$ if $\Delta(x, y) < \frac{n}{2} - \sqrt{n}$

Want: randomized, constant error protocol

Cost: Worst case number of bits communicated

$x =$	0	1	0	0	1	0	1	1	0	0	0	1
$y =$	0	0	0	0	0	0	1	1	1	0	0	1

$$n = 12; \quad \Delta(x, y) = 3 \in [6 - \sqrt{12}, 6 + \sqrt{12}]$$

Implications

Data stream lower bounds

- Distinct elements
- Frequency moments
- Norms
- Entropy
- General form of bound: $ps = \Omega(1/\varepsilon^2)$

Distributed functional monitoring lower bounds

Connections to differential privacy

The Reductions

E.g., Distinct Elements (Other problems: similar)

$x =$	0	1	0	0	1	0	1	1	0	0	0	1
$\sigma :$	$(1,0)$	$(2,1)$	$(3,0)$	$(4,0)$	$(5,1)$	$(6,0)$	$(7,1)$	$(8,1)$	$(9,0)$	$(10,0)$	$(11,0)$	$(12,1)$
$y =$	0	0	0	0	0	0	1	1	1	0	0	1
$\tau :$	$(1,0)$	$(2,0)$	$(3,0)$	$(4,0)$	$(5,0)$	$(6,0)$	$(7,1)$	$(8,1)$	$(9,1)$	$(10,0)$	$(11,0)$	$(12,1)$

Alice: $x \mapsto \sigma = \langle (1, x_1), (2, x_2), \dots, (n, x_n) \rangle$

Bob: $y \mapsto \tau = \langle (1, y_1), (2, y_2), \dots, (n, y_n) \rangle$

Notice: $F_0(\sigma \circ \tau) = n + \Delta(x, y) = \begin{cases} < \frac{3n}{2} - \sqrt{n}, & \text{or} \\ > \frac{3n}{2} + \sqrt{n}. \end{cases} \quad \text{Set } \varepsilon = \frac{1}{\sqrt{n}}.$

Ancient History



One-Pass Bounds

Indyk, Woodruff [FOCS 2003]

- Considered one-pass lower bound for DIST-ELEM
- Recognized relevance of GHD, difficulty of lower-bounding
- Defined “related” problem Π_{ℓ_2} , showed $R^{\rightarrow}(\Pi_{\ell_2}) = \Omega(n)$
- Concluded $\Omega(\varepsilon^{-2})$ bound for $\text{DIST-ELEM}_{m,\varepsilon}$ with $m = \tilde{\Omega}(1/\varepsilon^9)$

One-Pass Bounds

Indyk, Woodruff [FOCS 2003]

- Considered one-pass lower bound for DIST-ELEM
- Recognized relevance of GHD, difficulty of lower-bounding
- Defined “related” problem Π_{ℓ_2} , showed $R^{\rightarrow}(\Pi_{\ell_2}) = \Omega(n)$
- Concluded $\Omega(\varepsilon^{-2})$ bound for $\text{DIST-ELEM}_{m,\varepsilon}$ with $m = \tilde{\Omega}(1/\varepsilon^9)$

Woodruff [SODA 2004]

- Worked with GHD itself, showed $R^{\rightarrow}(\text{GHD}) = \Omega(n)$
- Very intricate combinatorial proof, with hairy probability estimations
- Conjectured $R(\text{GHD}) = \Omega(n)$, implying multi-pass lower bounds

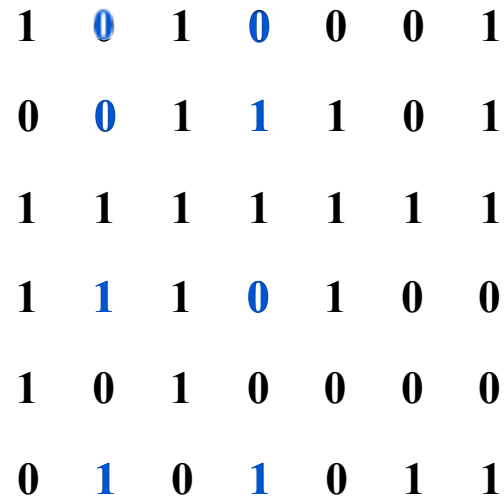
The VC-Dimension Technique

- Consider communication matrix of GHD as set system
- The system has $\Omega(n)$ VC-dimension

1	0	1	0	0	0	1
0	0	1	1	1	0	1
1	1	1	1	1	1	1
1	1	1	0	1	0	0
1	0	1	0	0	0	0
0	1	0	1	0	1	1

The VC-Dimension Technique

- Consider communication matrix of GHD as set system
- The system has $\Omega(n)$ VC-dimension

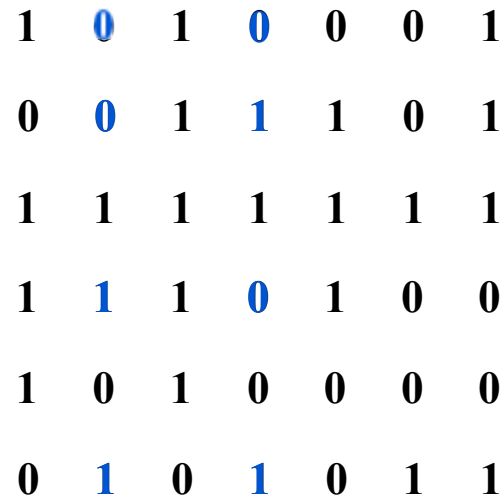


1	0	1	0	0	0	1
0	0	1	1	1	0	1
1	1	1	1	1	1	1
1	1	1	0	1	0	0
1	0	1	0	0	0	0
0	1	0	1	0	1	1

Instance of INDEX

The VC-Dimension Technique

- Consider communication matrix of GHD as set system
- The system has $\Omega(n)$ VC-dimension

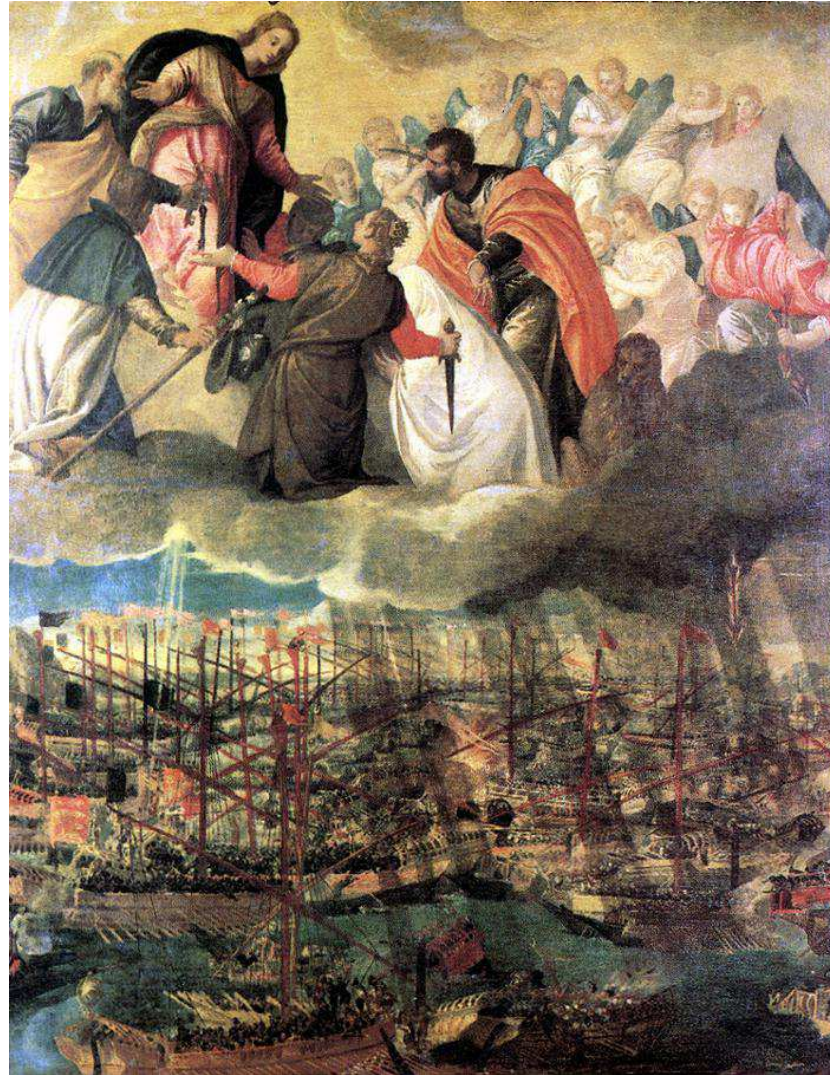


1	0	1	0	0	0	1
0	0	1	1	1	0	1
1	1	1	1	1	1	1
1	1	1	0	1	0	0
1	0	1	0	0	0	0
0	1	0	1	0	1	1

Instance of INDEX

- Thus, $R^{\rightarrow}(\text{GHD}) = \Omega(n)$

The Middle Ages



A Nice Simplification

Jayram, Kumar, Sivakumar [circa 2005]

- Simpler proof of $R^{\rightarrow}(\text{GHD}) = \Omega(n)$
- *Much* simpler: direct reduction from INDEX
- Geometric intuition:

$$\text{Alice: } x \in \{0, 1\}^n \longmapsto \tilde{x} \in \left\{ \frac{1}{\sqrt{n}}, -\frac{1}{\sqrt{n}} \right\}^n \in \mathbb{R}^n$$

$$\text{Bob: } j \in [n] \longmapsto e_j = (0, \dots, 0, 1, 0, \dots, 0) \in \mathbb{R}^n$$

- Observe: $\langle \tilde{x}, e_j \rangle \neq 0$, and x_j determined by $\text{sgn}\langle \tilde{x}, e_j \rangle$
- We've reduced INDEX to “gap-inner-product”, or GIP

Inner Product \leftrightarrow Hamming Distance

- Obviously, GHD \rightarrow GIP:

$$\langle \tilde{x}, \tilde{y} \rangle = 1 - \frac{2\Delta(x, y)}{n}$$

$$\langle \tilde{x}, \tilde{y} \rangle \gtrless \mp \frac{2}{\sqrt{n}} \Rightarrow \Delta(x, y) \lesseqgtr \frac{n}{2} \pm \sqrt{n}$$

- Also, GIP \rightarrow GHD by “discretization transform”:

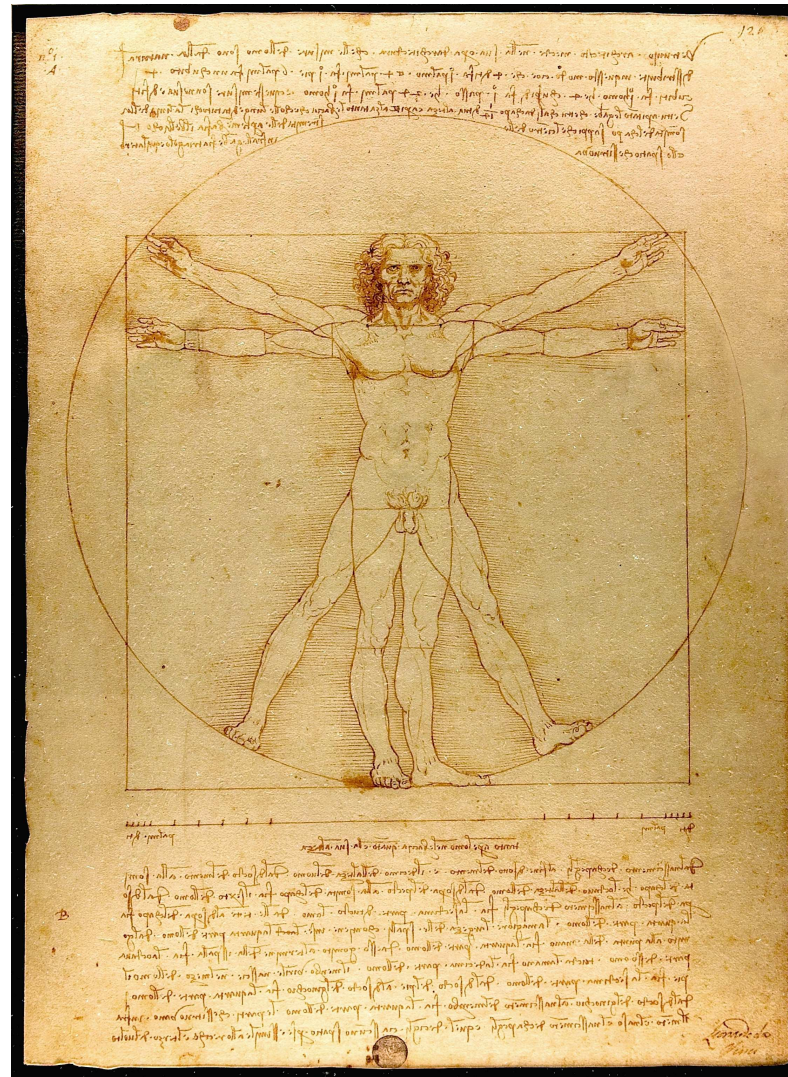
Pick random Gaussians r_1, \dots, r_N , with $N = 10n$

Alice: $\bar{x} \in \mathbb{R}^n \mapsto x = (\text{sgn}\langle \bar{x}, r_1 \rangle, \dots, \text{sgn}\langle \bar{x}, r_N \rangle) \in \{\pm 1\}^N$

Bob: $\bar{y} \in \mathbb{R}^n \mapsto y = (\text{sgn}\langle \bar{y}, r_1 \rangle, \dots, \text{sgn}\langle \bar{y}, r_N \rangle) \in \{\pm 1\}^N$

$$\langle \bar{x}, \bar{y} \rangle \gtrless \mp \frac{1}{\sqrt{n}} \xrightarrow{\text{whp}} \Delta(x, y) \lesseqgtr \frac{N}{2} \pm O(\sqrt{N})$$

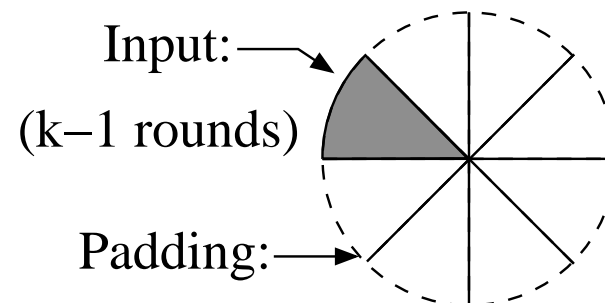
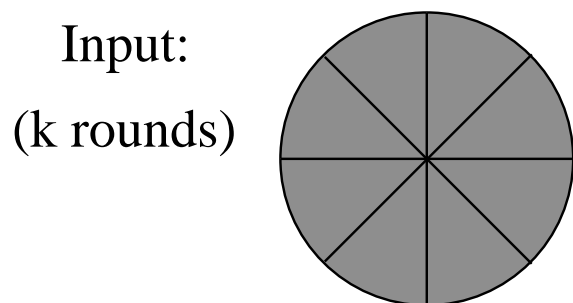
The Renaissance Era



Round Elimination

Brody, Chakrabarti [CCC 2009]


- Can we at least rule out a *two-pass* improvement for DIST-ELEM?
- A cheap first message makes little progress? Then rinse, repeat
- Tends to decimate problem [Miltersen-Nisan-Safra-Wigderson'98] [Sen'03]



Another VC-Dimension Argument: Subcube Lifting

First message constant on large set:

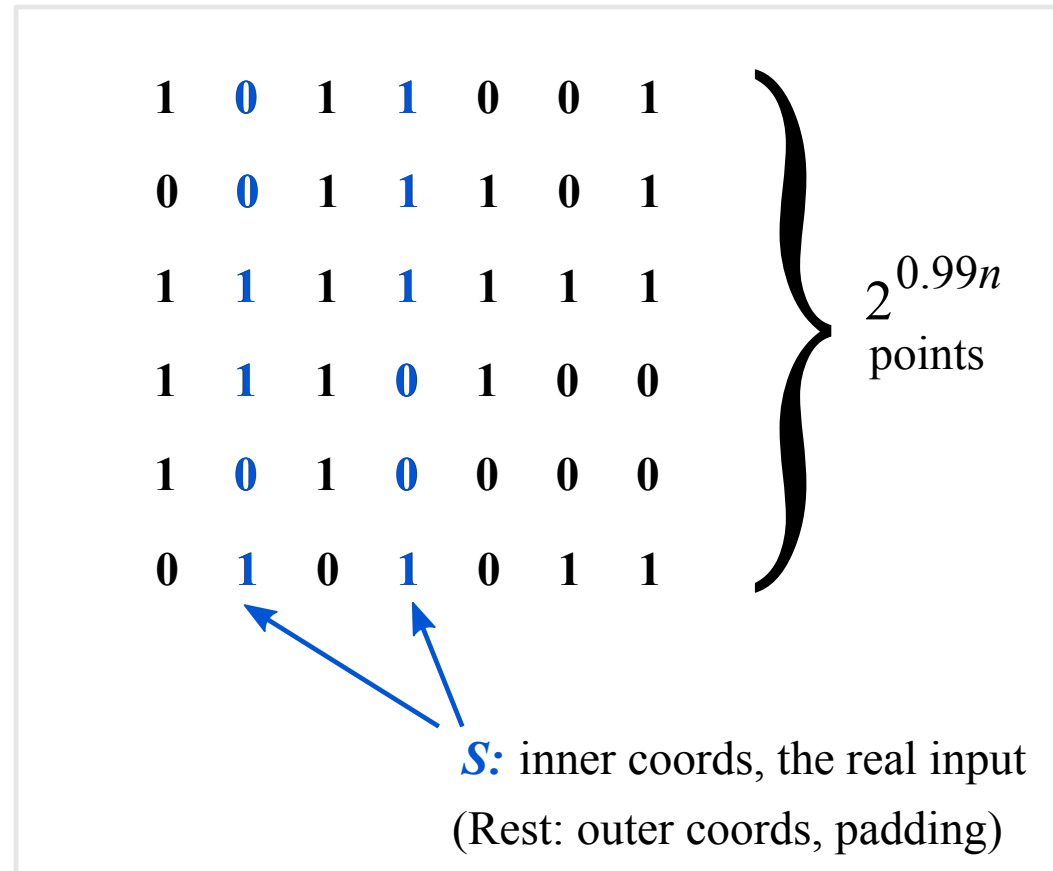
1	0	1	1	0	0	1
0	0	1	1	1	0	1
1	1	1	1	1	1	1
1	1	1	0	1	0	0
1	0	1	0	0	0	0
0	1	0	1	0	1	1


$$2^{0.99n}$$

points

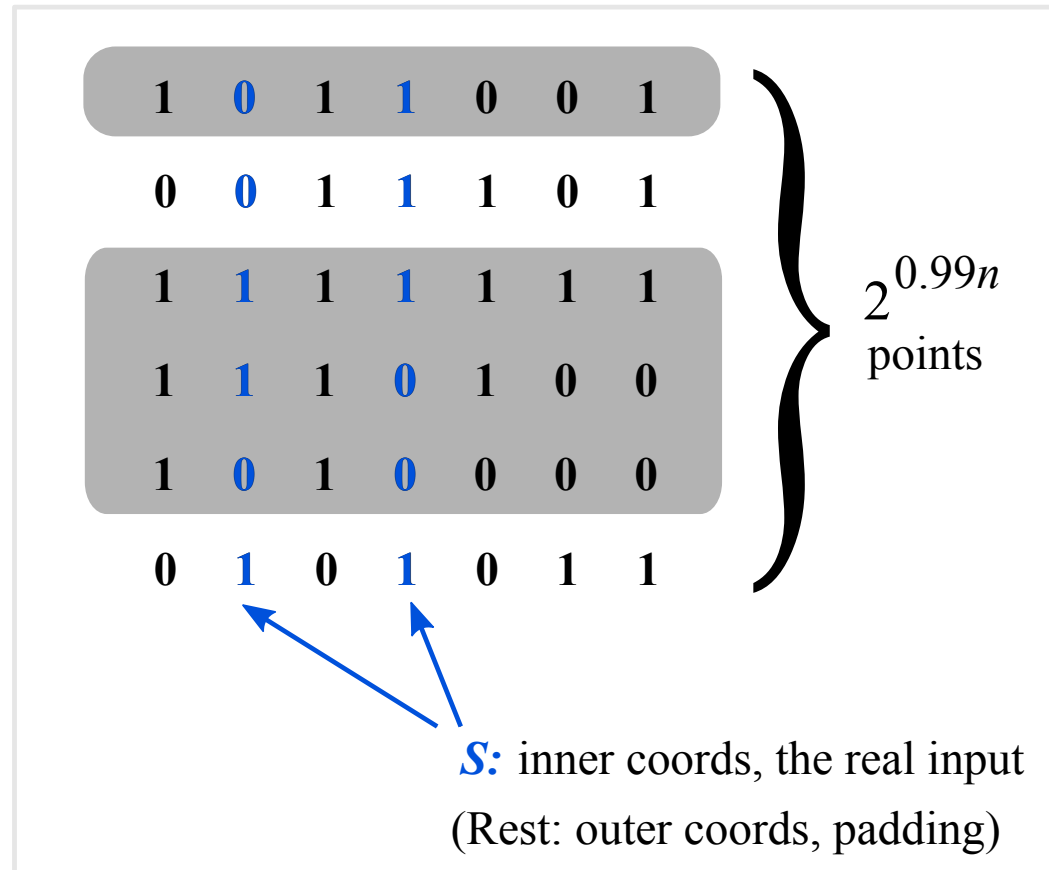
Another VC-Dimension Argument: Subcube Lifting

First message constant on large set:



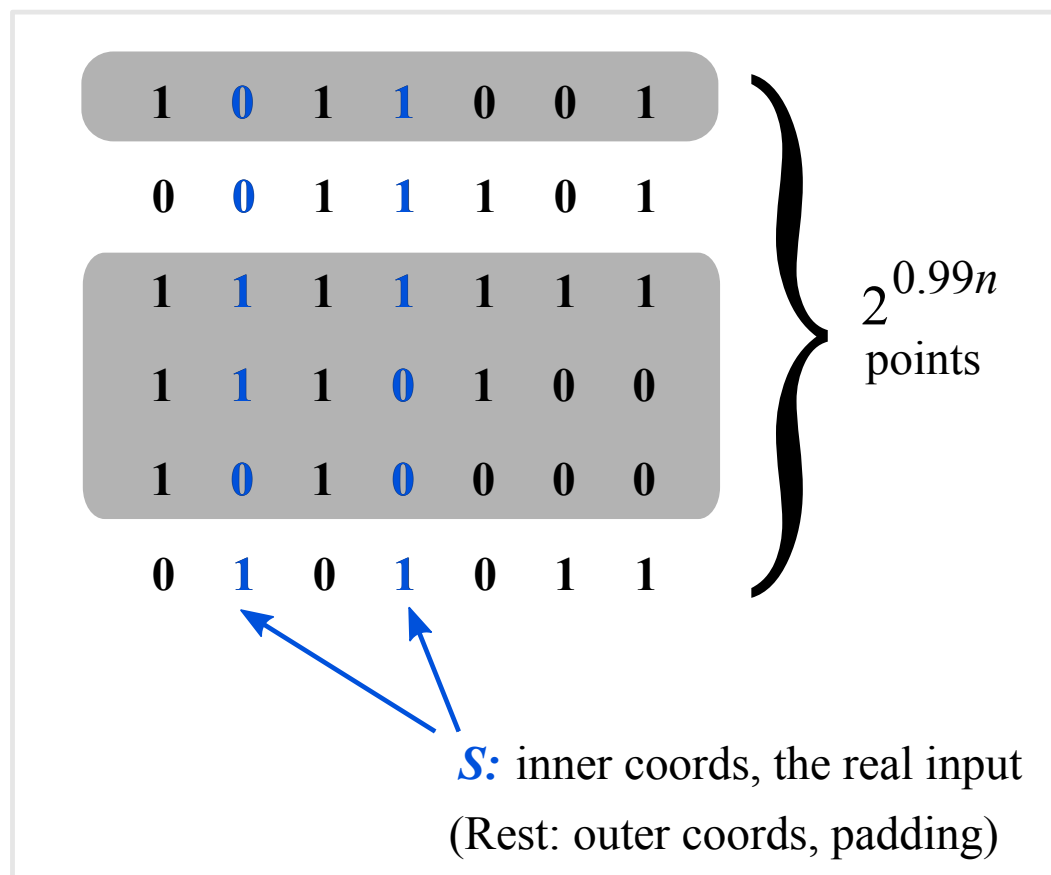
Another VC-Dimension Argument: Subcube Lifting

First message constant on large set:



Another VC-Dimension Argument: Subcube Lifting

First message constant on large set:



Alice, Bob lift their $(n/3)$ -dim inputs from **inner coords** to full n -dim space

First message now redundant, so eliminate!

[Brody-C.'09]

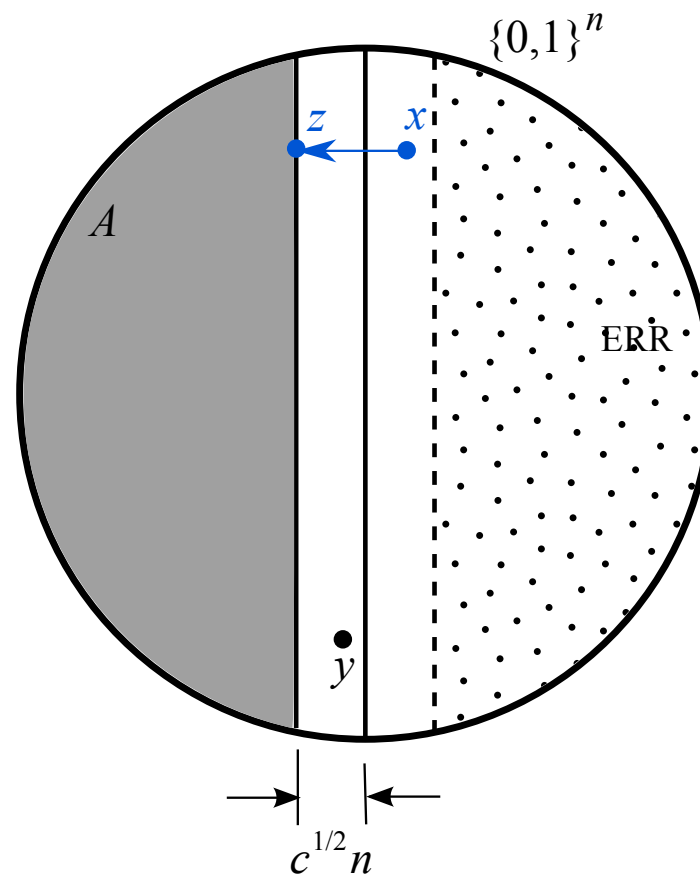
Better Round Elimination

Brody, Chakrabarti, Regev, Vidick, de Wolf [RANDOM 2010]

- Previous argument reduced dimension too rapidly
- Gives $R^k(\text{GHD}) = n/2^{O(k^2)}$
- Can improve to $R^k(\text{GHD}) = n/O(k^2)$

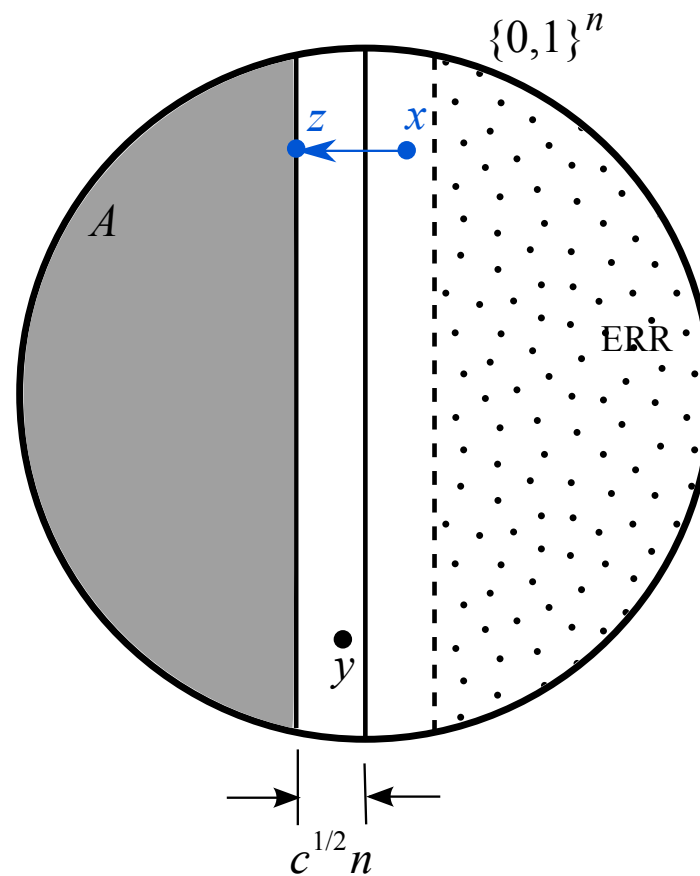
Round Elimination V2.0: Geometric Perturbation

First message constant over large set A



Round Elimination V2.0: Geometric Perturbation

First message constant over large set A



Alice: replace x with $z = \text{NearestNeighbour}(x, A)$

Modern History



Main Theorem

Chakrabarti, Regev [STOC 2011]

And now, we show:

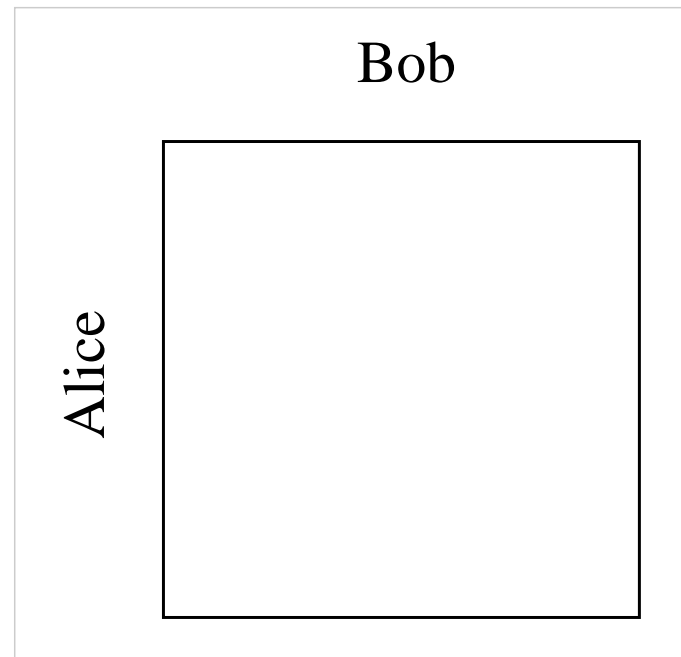
$$R(\text{GHD}) = \Omega(n)$$

The Rectangle Property

Input universe $U = \{0, 1\}^n \times \{0, 1\}^n$

Deterministic protocol P , communicating $\leq c$ bits

partitions U into $\leq 2^c$ rectangles $A_i \times B_i$, where $A_i, B_i \subseteq \{0, 1\}^n$

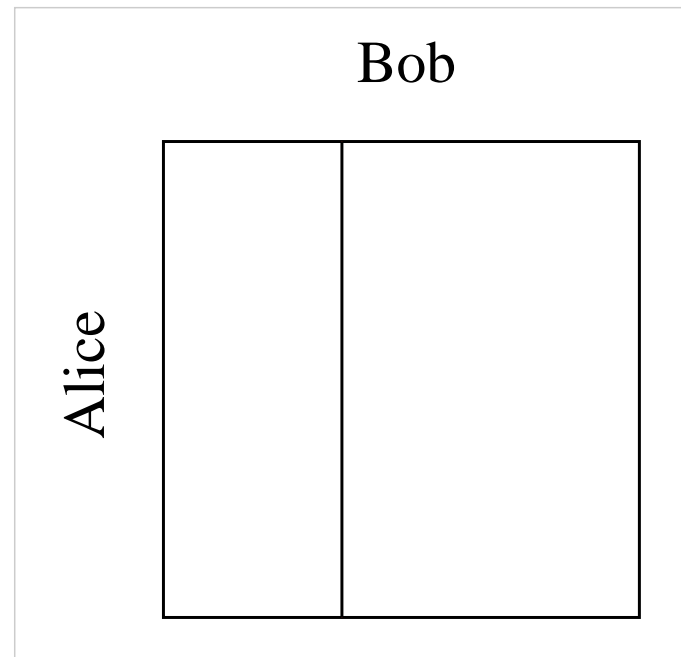


The Rectangle Property

Input universe $U = \{0, 1\}^n \times \{0, 1\}^n$

Deterministic protocol P , communicating $\leq c$ bits

partitions U into $\leq 2^c$ rectangles $A_i \times B_i$, where $A_i, B_i \subseteq \{0, 1\}^n$

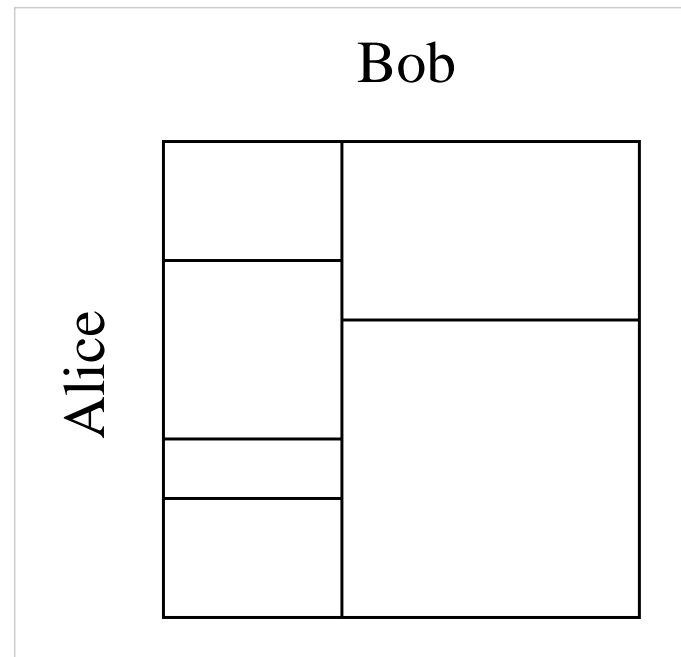


The Rectangle Property

Input universe $U = \{0, 1\}^n \times \{0, 1\}^n$

Deterministic protocol P , communicating $\leq c$ bits

partitions U into $\leq 2^c$ rectangles $A_i \times B_i$, where $A_i, B_i \subseteq \{0, 1\}^n$

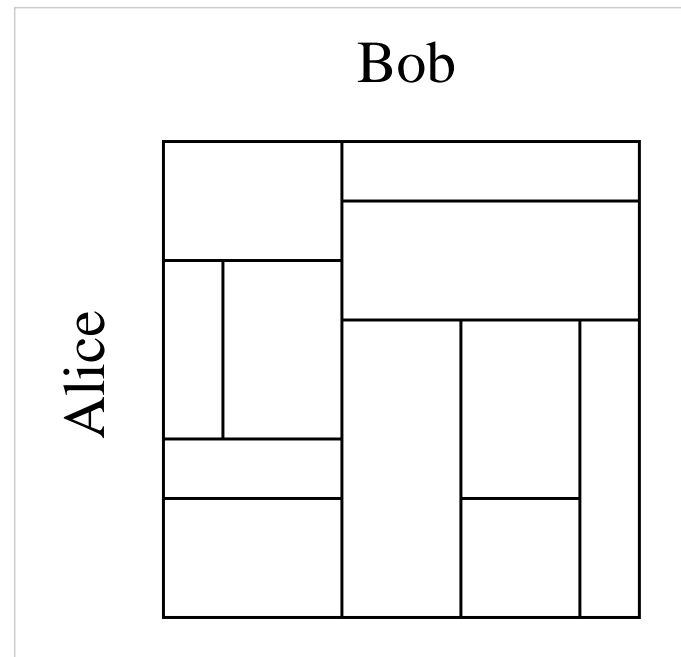


The Rectangle Property

Input universe $U = \{0, 1\}^n \times \{0, 1\}^n$

Deterministic protocol P , communicating $\leq c$ bits

partitions U into $\leq 2^c$ rectangles $A_i \times B_i$, where $A_i, B_i \subseteq \{0, 1\}^n$

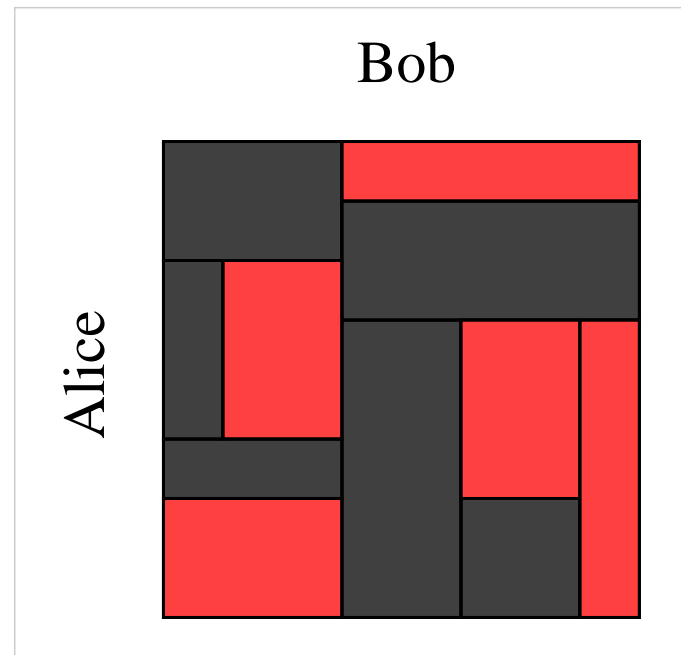


The Rectangle Property

Input universe $U = \{0, 1\}^n \times \{0, 1\}^n$

Deterministic protocol P , communicating $\leq c$ bits

partitions U into $\leq 2^c$ rectangles $A_i \times B_i$, where $A_i, B_i \subseteq \{0, 1\}^n$

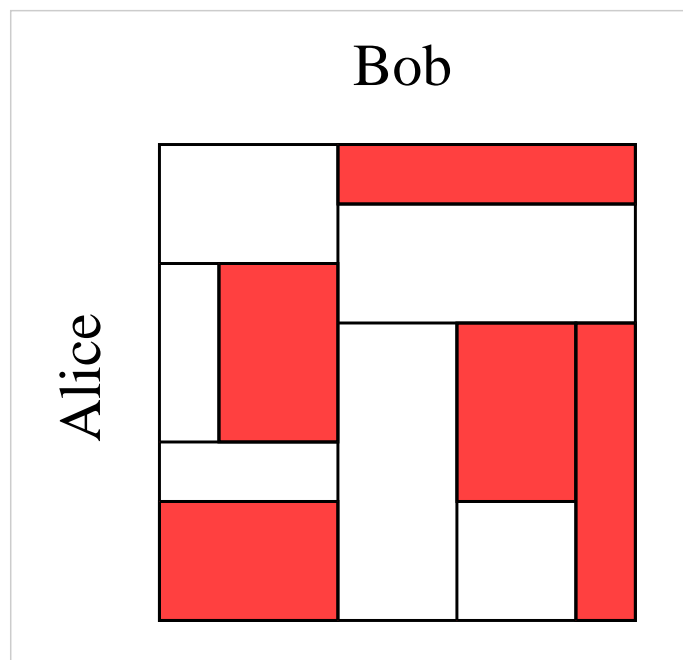


The Rectangle Property

Input universe $U = \{0, 1\}^n \times \{0, 1\}^n$

Deterministic protocol P , communicating $\leq c$ bits

partitions U into $\leq 2^c$ rectangles $A_i \times B_i$, where $A_i, B_i \subseteq \{0, 1\}^n$



If P computes $f : U \rightarrow \{0, 1\}$, then $f^{-1}(0) = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

The Corruption Technique and a Twist

Deterministic: $f^{-1}(0) = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

Randomized: $\{P \text{ outputs } 0\} = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

- Partition covers *most of* $f^{-1}(0)$
- Each R_i mostly *uncorrupted*: contains much fewer 1s than 0s.

The Corruption Technique and a Twist

Deterministic: $f^{-1}(0) = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

Randomized: $\{P \text{ outputs } 0\} = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

- Partition covers *most of* $f^{-1}(0)$
- Each R_i mostly *uncorrupted*: contains much fewer 1s than 0s.

For lower bound:

- Show every large rectangle (size $\geq 2^{0.99n} \times 2^{0.99n}$) is *corrupted*

$$\mu_1(R) \geq \alpha \mu_0(R)$$

The Corruption Technique and a Twist

Deterministic: $f^{-1}(0) = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

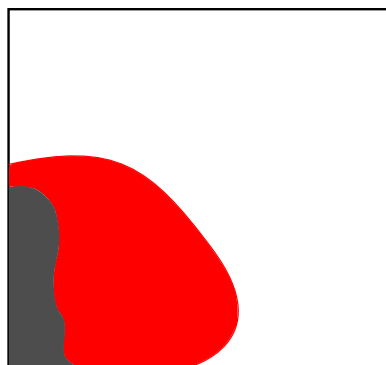
Randomized: $\{P \text{ outputs } 0\} = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

- Partition covers *most* of $f^{-1}(0)$
- Each R_i mostly *uncorrupted*: contains much fewer 1s than 0s.

For lower bound:

- Show every large rectangle (size $\geq 2^{0.99n} \times 2^{0.99n}$) is *corrupted*

$$\mu_1(R) \geq \alpha \mu_0(R)$$



The Corruption Technique and a Twist

Deterministic: $f^{-1}(0) = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

Randomized: $\{P \text{ outputs } 0\} = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

- Partition covers *most* of $f^{-1}(0)$
- Each R_i mostly *uncorrupted*: contains much fewer 1s than 0s.

For lower bound:

- Show every large rectangle (size $\geq 2^{0.99n} \times 2^{0.99n}$) is *corrupted*

$$\mu_1(R) \geq \alpha \mu_0(R)$$

- Caveat: not true! E.g., $\{(x, y) : x_{1:100\sqrt{n}} = y_{1:100\sqrt{n}} = \vec{0}\}$

The Corruption Technique and a Twist

Deterministic: $f^{-1}(0) = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

Randomized: $\{P \text{ outputs } 0\} = R_1 \cup R_2 \cup \dots \cup R_{2^c}$

- Partition covers *most* of $f^{-1}(0)$
- Each R_i mostly *uncorrupted*: contains much fewer 1s than 0s.

For lower bound:

- Show every large rectangle (size $\geq 2^{0.99n} \times 2^{0.99n}$) is *corrupted*

$$\mu_1(R) \geq \alpha \mu_0(R)$$

- Caveat: not true! E.g., $\{(x, y) : x_{1:100\sqrt{n}} = y_{1:100\sqrt{n}} = \vec{0}\}$
- Show weaker inequality

$$\mu_1(R) + \beta \mu_\star(R) \geq \alpha \mu_0(R) \quad (\alpha > \beta)$$

Corruption with Jokers

Pick distribs μ_0, μ_1 on $f^{-1}(0), f^{-1}(1)$, and another distrib μ_\star

Argue that for all large rectangles R , we have

$$\mu_1(R) + \beta \mu_\star(R) \geq \alpha \mu_0(R) \quad (\alpha > \beta)$$



Corruption with Jokers

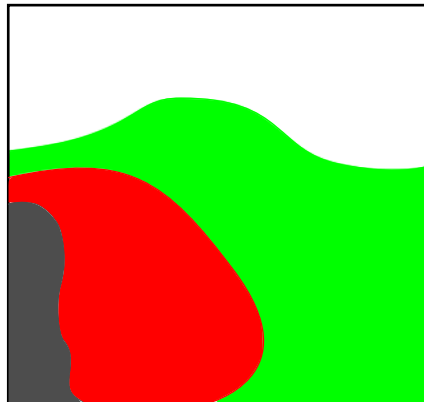
Pick distribs μ_0, μ_1 on $f^{-1}(0), f^{-1}(1)$, and another distrib μ_\star

Argue that for all large rectangles R , we have

$$\mu_1(R) + \beta \mu_\star(R) \geq \alpha \mu_0(R) \quad (\alpha > \beta)$$

Sum over partition $\{P \text{ outputs } 0\} = \bigcup_{i=1}^{2^c} R_i$:

$$\mu_1(P^{-1}(0)) + \beta \mu_\star(P^{-1}(0)) \geq \alpha \mu_0(P^{-1}(0))$$



Corruption with Jokers

Pick distribs μ_0, μ_1 on $f^{-1}(0), f^{-1}(1)$, and another distrib μ_\star

Argue that for all large rectangles R , we have

$$\mu_1(R) + \beta \mu_\star(R) \geq \alpha \mu_0(R) \quad (\alpha > \beta)$$

Sum over partition $\{P \text{ outputs } 0\} = \bigcup_{i=1}^{2^c} R_i$:

$$\mu_1(P^{-1}(0)) + \beta \mu_\star(P^{-1}(0)) \geq \alpha \mu_0(P^{-1}(0)) \geq \alpha(1 - \varepsilon)$$



Corruption with Jokers

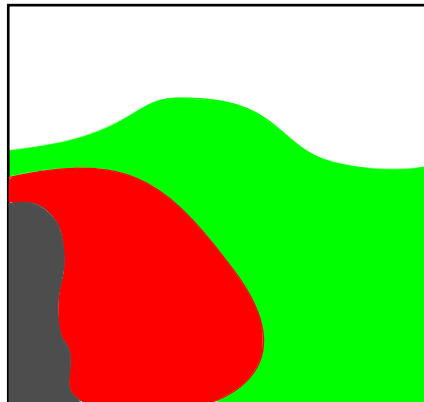
Pick distribs μ_0, μ_1 on $f^{-1}(0), f^{-1}(1)$, and another distrib μ_\star

Argue that for all large rectangles R , we have

$$\mu_1(R) + \beta \mu_\star(R) \geq \alpha \mu_0(R) \quad (\alpha > \beta)$$

Sum over partition $\{P \text{ outputs } 0\} = \bigcup_{i=1}^{2^c} R_i$:

$$\varepsilon + \beta \geq \mu_1(P^{-1}(0)) + \beta \mu_\star(P^{-1}(0)) \geq \alpha \mu_0(P^{-1}(0)) \geq \alpha(1 - \varepsilon)$$



The Corruption Inequality and Its Proof

Let

$$\mu_0 = \text{Uniform on } \{(x, y) : \langle \tilde{x}, \tilde{y} \rangle = 0\}$$

$$\mu_1 = \text{Uniform on } \{(x, y) : \langle \tilde{x}, \tilde{y} \rangle = -10/\sqrt{n}\}$$

$$\mu_\star = \text{Uniform on } \{(x, y) : \langle \tilde{x}, \tilde{y} \rangle = 10/\sqrt{n}\}$$

The Key Inequality: For $|A|, |B| \geq 2^{0.99n}$

$$\frac{1}{2}(\mu_1(A \times B) + \mu_\star(A \times B)) \geq \frac{9}{10} \mu_0(A \times B)$$

“Inner product between large sets not too concentrated around zero”

The Corruption Inequality and Its Proof

Let

$$\mu_0 = \text{Uniform on } \{(x, y) : \langle \tilde{x}, \tilde{y} \rangle = 0\}$$

$$\mu_1 = \text{Uniform on } \{(x, y) : \langle \tilde{x}, \tilde{y} \rangle = -10/\sqrt{n}\}$$

$$\mu_\star = \text{Uniform on } \{(x, y) : \langle \tilde{x}, \tilde{y} \rangle = 10/\sqrt{n}\}$$

The Key Inequality: For $|A|, |B| \geq 2^{0.99n}$

$$\frac{1}{2}(\mu_1(A \times B) + \mu_\star(A \times B)) \geq \frac{9}{10} \mu_0(A \times B)$$

“Inner product between large sets not too concentrated around zero”

Proof Strategy: For $A, B \subseteq \mathbb{R}^n$ with $\gamma(A), \gamma(B) \geq 2^{-0.01n}$

distrib of $\langle \hat{x}, \hat{y} \rangle$ “spread out” like $N(0, 1)$

where $\gamma = n$ -dim Gaussian, $(\hat{x}, \hat{y}) \leftarrow A \times B$

Proof Details

Goal: For $A, B \subseteq \mathbb{R}^n$ with $\gamma(A), \gamma(B) \geq 2^{-0.01n}$

distrib of $\langle \hat{x}, \hat{y} \rangle$ “spread out” like $N(0, 1)$

Think

$A = \{\text{directions}\}$

$A_{\text{bad}} = \{\text{bad directions in } A\}$

$= \{\hat{x} \in A : \langle \hat{x}, \hat{y} \rangle \text{ not spread out, for } \hat{y} \leftarrow B\}$

Proof Details

Goal: For $A, B \subseteq \mathbb{R}^n$ with $\gamma(A), \gamma(B) \geq 2^{-0.01n}$

distrib of $\langle \hat{x}, \hat{y} \rangle$ “spread out” like $N(0, 1)$

Think

$$A = \{\text{directions}\}$$

$$A_{\text{bad}} = \{\text{bad directions in } A\}$$

$$= \{\hat{x} \in A : \langle \hat{x}, \hat{y} \rangle \text{ not spread out, for } \hat{y} \leftarrow B\}$$

For a contradiction, suppose $\gamma(A_{\text{bad}}) > 2^{-0.02n}$

Then (Raz’s Lemma): A contains orthogonal bad dirs $\hat{x}_1, \dots, \hat{x}_{n/2}$

Proof Details

Goal: For $A, B \subseteq \mathbb{R}^n$ with $\gamma(A), \gamma(B) \geq 2^{-0.01n}$

distrib of $\langle \hat{x}, \hat{y} \rangle$ “spread out” like $N(0, 1)$

Think

$$\begin{aligned} A &= \{\text{directions}\} \\ A_{\text{bad}} &= \{\text{bad directions in } A\} \\ &= \{\hat{x} \in A : \langle \hat{x}, \hat{y} \rangle \text{ not spread out, for } \hat{y} \leftarrow B\} \end{aligned}$$

For a contradiction, suppose $\gamma(A_{\text{bad}}) > 2^{-0.02n}$

Then (Raz’s Lemma): A contains orthogonal bad dirs $\hat{x}_1, \dots, \hat{x}_{n/2}$

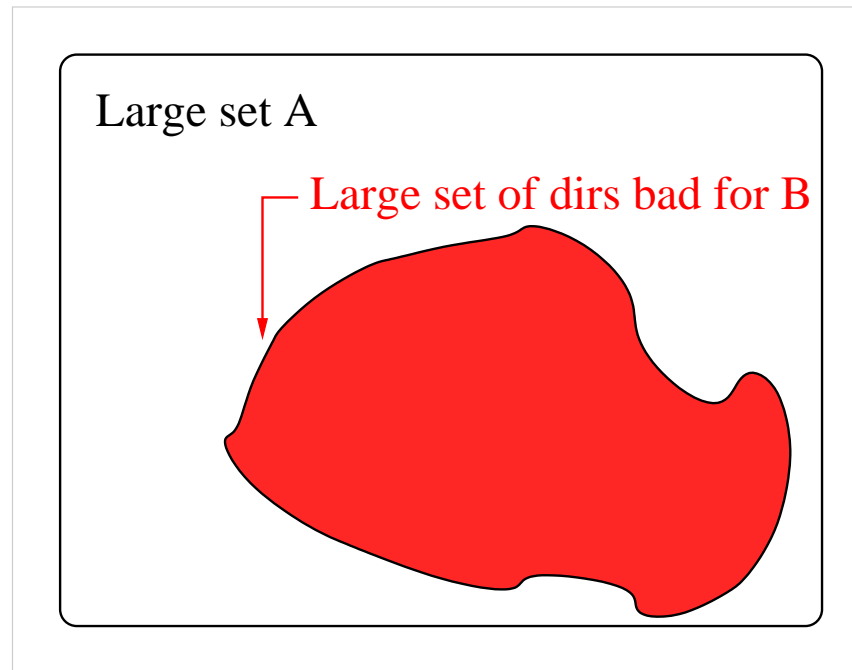
Therefore (Information Theory): $\hat{y} \leftarrow B$ can’t have enough entropy

Contradicts $\gamma(B) \geq 2^{-0.01n}$

Geometric and Info Theoretic Intuition

Large set A

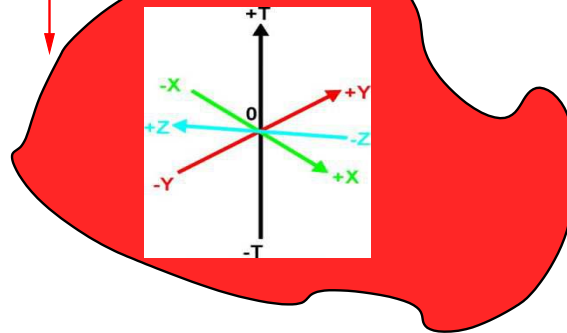
Geometric and Info Theoretic Intuition



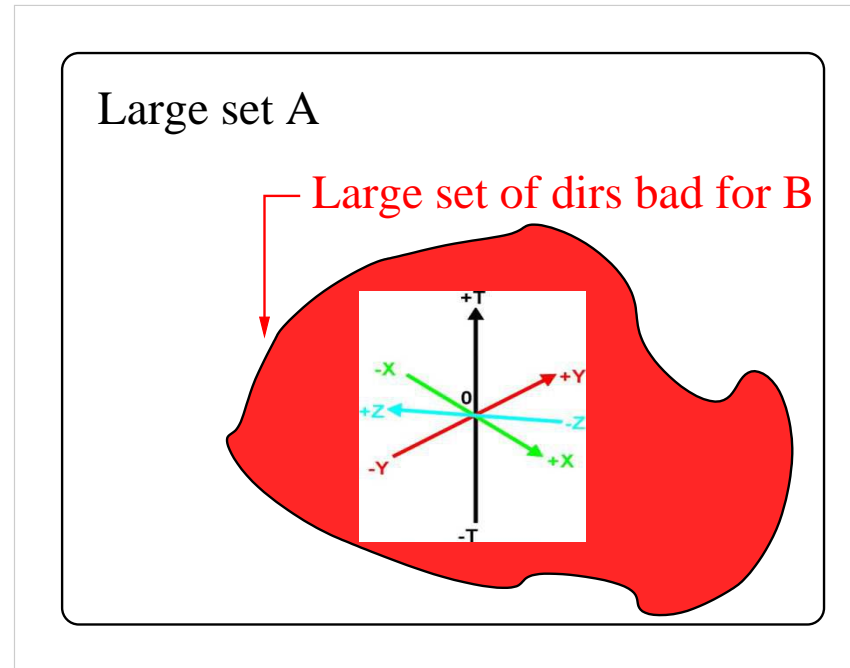
Geometric and Info Theoretic Intuition

Large set A

Large set of dirs bad for B



Geometric and Info Theoretic Intuition



$$\begin{aligned}
 0.99n &\leq H(y) \leq H(\langle y, x_1 \rangle, \dots, \langle y, x_n \rangle) \\
 &= \sum_{k=1}^{n/2} H(\langle y, x_k \rangle \mid \langle y, x_1 \rangle, \dots, \langle y, x_{k-1} \rangle) \\
 &\quad + \sum_{k=n/2+1}^n H(\langle y, x_k \rangle \mid \langle y, x_1 \rangle, \dots, \langle y, x_{k-1} \rangle) \\
 &\leq \sum_{k=1}^{n/2} 0.7 + \sum_{k=n/2+1}^n 1 = 0.85n
 \end{aligned}$$

The Future



The Future

Two simplifications of our proof [not yet published]

- Vidick shows following anti-concentration inequality:

$$\mathbb{E}[\langle \tilde{x}, \tilde{y} \rangle^2] = \Omega(1/n)$$

Avoids “continuous information theory”; just concentration of measure

- Sherstov: anti-concentration gives corruption-based proof that

$$R(\text{NEAR-ORTHOGONAL}) = \Omega(n)$$

and reduces NEAR-ORTHOGONAL to GHD; thus avoids “jokers”

- Also, Sherstov proves anti-concentration using Talagrand’s inequality

Conclusions

- Settled communication complexity of GHD, proving a long-conjectured $\Omega(n)$ bound
- As a result, understood multi-pass space complexity of a number of data stream problems

Conclusions

- Settled communication complexity of GHD, proving a long-conjectured $\Omega(n)$ bound
- As a result, understood multi-pass space complexity of a number of data stream problems

Open Problem

Prove that GHD is hard under the uniform distribution