

APPROXIMATE DEGREE AND QUANTUM QUERY LOWER BOUNDS VIA DUAL POLYNOMIALS

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Approximate Degree [Nisan-Szegedy92]

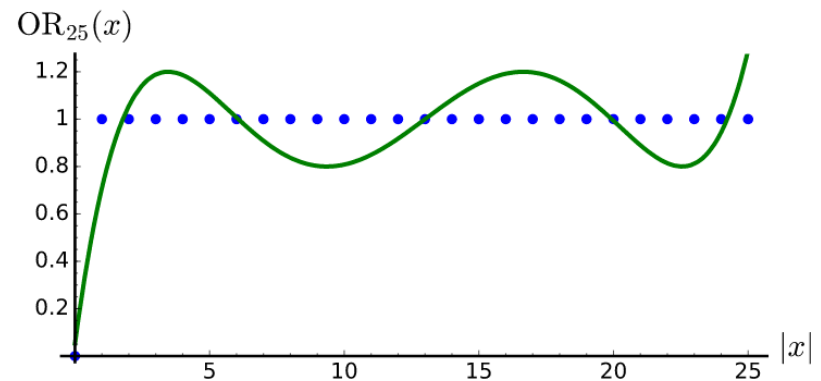
For a Boolean function $f : \{0, 1\}^n \rightarrow \{0, 1\}$

Approximate Degree: Minimum degree of a real polynomial $p : \{0, 1\}^n \rightarrow \mathbb{R}$ such that

$$|p(x) - f(x)| \leq 1/3 \quad \text{for all } x \in \{0, 1\}^n$$

Denoted by $\text{adeg}(f)$

Ex. $\text{adeg}(\text{OR}_n) = \Theta(\sqrt{n})$



Research Directions for the Polynomial Method

(1) Advance our understanding of adeg

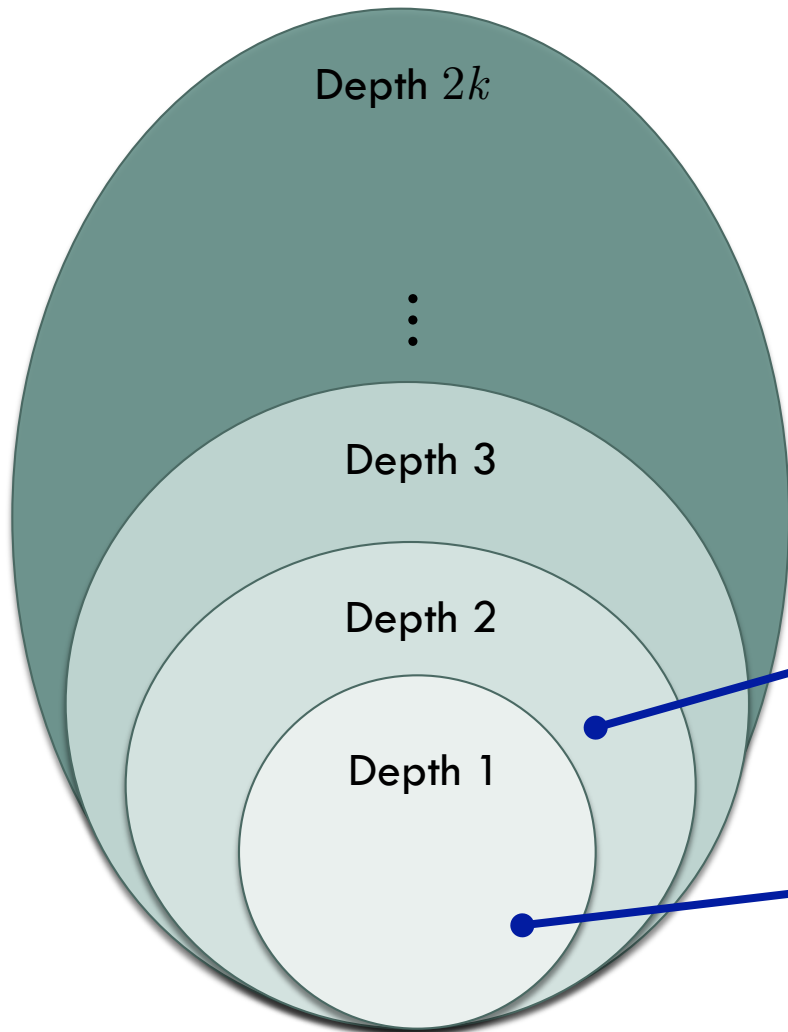
A Nearly Optimal Lower Bound on the Approximate Degree of AC^0

Hardness amplification within AC^0

(2) Use adeg to advance application domains

The Polynomial Method Strikes Back:
Tight Quantum Query Bounds via Dual Polynomials

Approximate Degree of AC^0

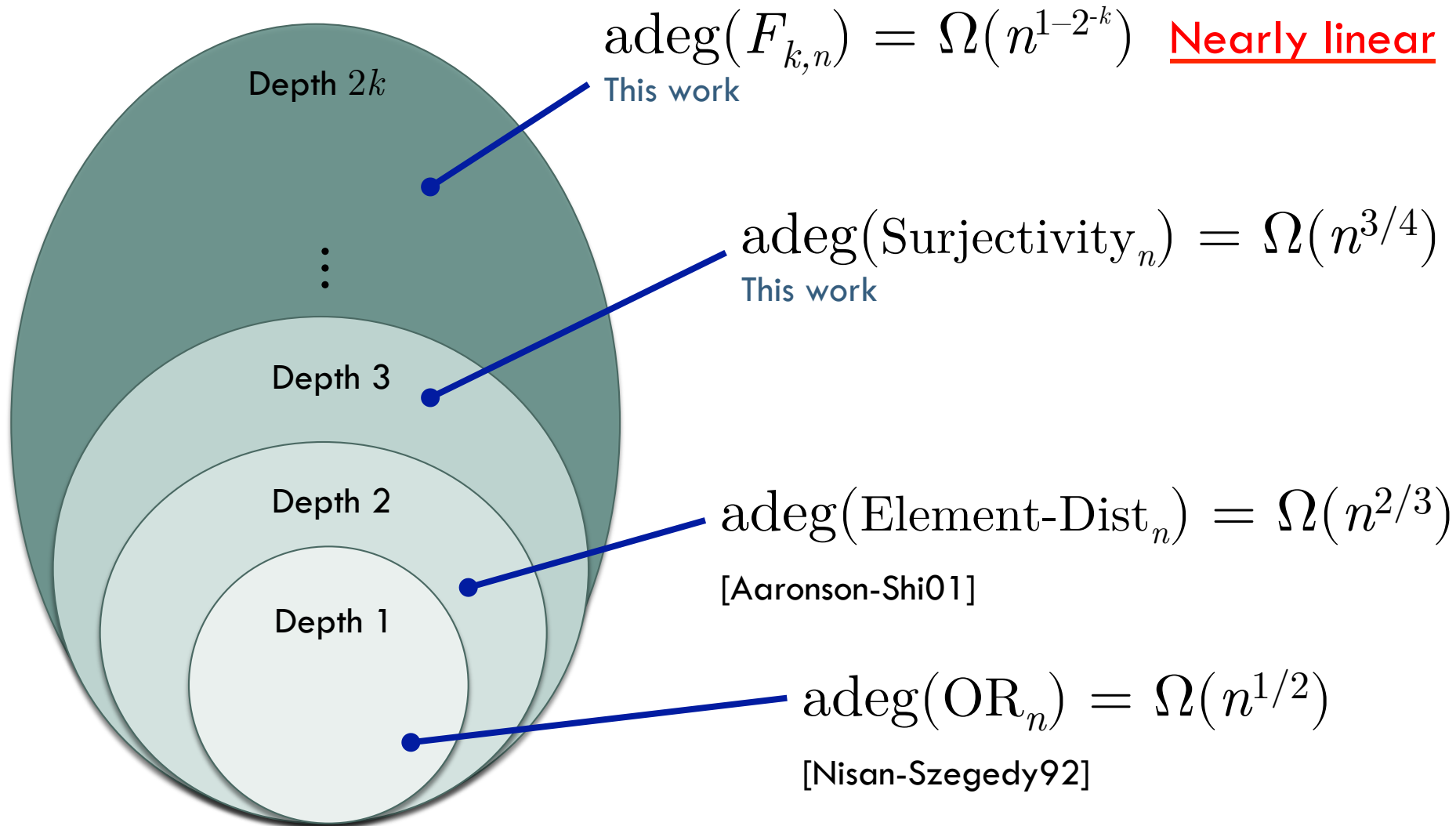


Q: Is there an AC^0 function with approximate degree $\Omega(n)$?

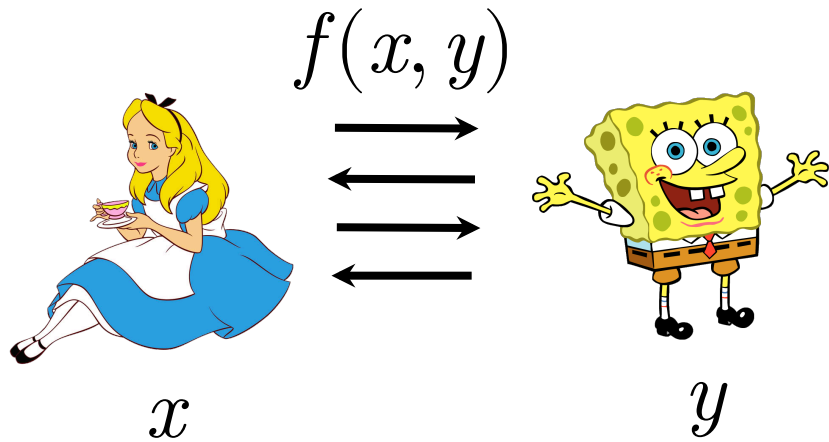
$\text{adeg}(\text{Element-Dist}_n) = \Omega(n^{2/3})$
[Aaronson-Shi01]

$\text{adeg}(\text{OR}_n) = \Omega(n^{1/2})$
[Nisan-Szegedy92]

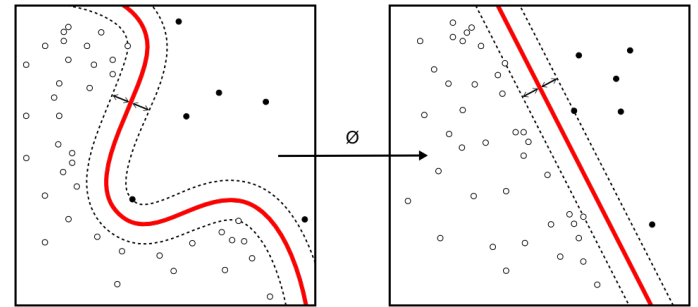
Approximate Degree of AC^0



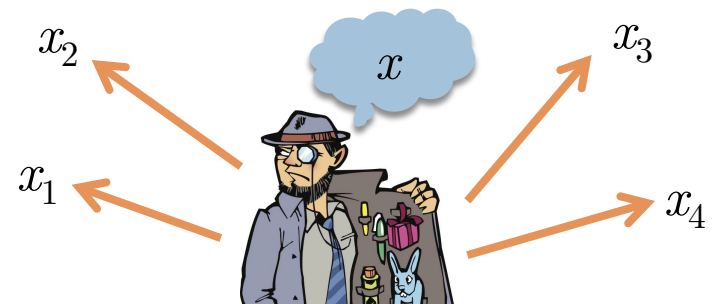
Applications of AC^0 Lower Bound



Nearly optimal $\Omega(n^{1-\delta})$
quantum and multiparty
communication lower bounds for AC^0



Learning via regression requires
 $\exp(\Omega(n^{1-\delta}))$ features



Improved secret sharing
with reconstruction in AC^0

Research Directions for the Polynomial Method

(1) Advance our understanding of adeg

A Nearly Optimal Lower Bound on the Approximate Degree of AC^0

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The Polynomial Method Strikes Back:
Tight Quantum Query Bounds via Dual Polynomials

(Deterministic) Query Complexity

Let $f : \{0, 1\}^n \rightarrow \{0, 1\}$ be a boolean function

Deterministic Query Complexity:

Minimum number of bits of x that must be read to compute $f(x)$

Ex. Computing OR_n requires n queries

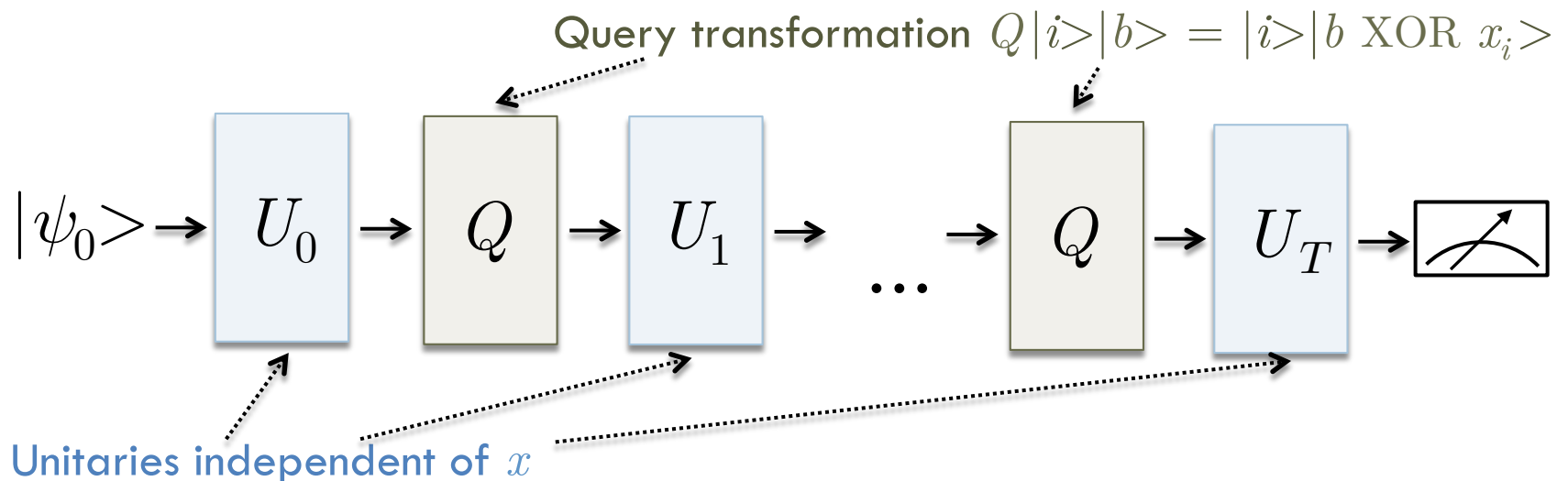
x_1	x_2	x_3	x_4	...	x_{n-1}	x_n

Quantum Query Complexity

Let $f : \{0, 1\}^n \rightarrow \{0, 1\}$ be a boolean function

Quantum Query Complexity:

Minimum number of bits of x that must be read **in superposition** to compute $f(x)$ **with probability $\geq 2/3$**



Quantum Query Complexity

Let $f : \{0, 1\}^n \rightarrow \{0, 1\}$ be a boolean function

Quantum Query Complexity:

Minimum number of bits of x that must be read **in superposition** to compute $f(x)$ **with probability $\geq 2/3$**

Ex. Computing OR_n

only needs \sqrt{n} quantum queries [Grover96]

Quantum Query Lower Bounds

“The Polynomial Method” [Beals-Buhrman-Cleve-Mosca-deWolf98]:

Accept prob. of a T query algorithm = Degree $2T$ polynomial in x

$$\Rightarrow \text{Quantum-query-complexity}(f) \geq \frac{1}{2} \text{adeg}(f)$$

Newer “adversary” methods:

□ Positive-weights method [Ambainis02]

Easy to apply, but limited in power

□ Negative-weights method [Høyer-Lee-Špalek07, ..., Reichardt11]

Tight characterization, but difficult to apply



This work: New and nearly tight quantum query lower bounds via the polynomial method

Our Results

Problem	Best Prior Upper Bound	Our Lower Bound	Best Prior Lower Bound
k -distinctness	$O(n^{3/4-1/(2^{k+2}-4)})$ [Bel12a]	$\tilde{\Omega}(n^{3/4-1/(2^k)})$	$\tilde{\Omega}(n^{2/3})$ [AS04]
Image Size Testing	$O(\sqrt{n} \log n)$ [ABRdW16]	$\tilde{\Omega}(\sqrt{n})$	$\tilde{\Omega}(n^{1/3})$ [ABRdW16]
k -junta Testing	$O(\sqrt{k} \log k)$ [ABRdW16]	$\tilde{\Omega}(\sqrt{k})$	$\tilde{\Omega}(k^{1/3})$ [ABRdW16]
SDU	$O(\sqrt{n})$ [BHH11]	$\tilde{\Omega}(\sqrt{n})$	$\tilde{\Omega}(n^{1/3})$ [BHH11, AS04]
Shannon Entropy	$\tilde{O}(\sqrt{n})$ [BHH11, LW17]	$\tilde{\Omega}(\sqrt{n})$	$\tilde{\Omega}(n^{1/3})$ [LW17]

Table 1: Our lower bounds on quantum query complexity and approximate degree vs. prior work.

Problem	Best Prior Upper Bound	Our Upper Bound	Our Lower Bound	Best Prior Lower Bound
Surjectivity	$\tilde{O}(n^{3/4})$ [She18]	$\tilde{O}(n^{3/4})$	$\tilde{\Omega}(n^{3/4})$	$\tilde{\Omega}(n^{2/3})$ [AS04]

Table 2: Our bounds on the approximate degree of Surjectivity vs. prior work.

Lower Bound for k -distinctness

Define k -DIST $_{N,R} : \{1, \dots, R\}^N \rightarrow \{0, 1\}$ by

$$k\text{-DIST}_{N,R}(s_1, \dots, s_N) = 1 \quad \text{iff} \\ \text{Some } r \in [R] \text{ appears } \geq k \text{ times in the input list}$$

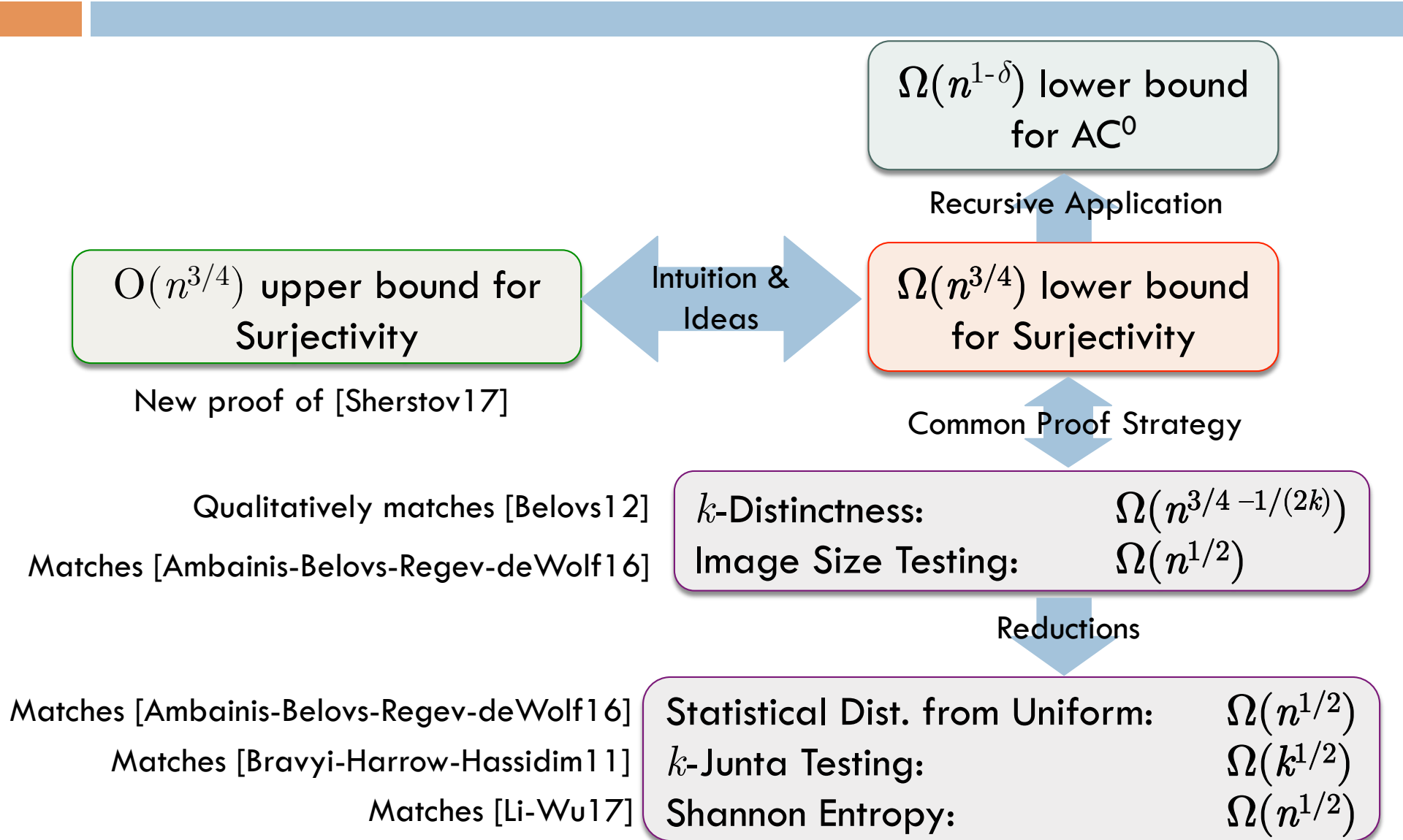
Corresponds to a Boolean function on $O(N \log_2 R)$ bits

Upper Bounds: $O(N^{k/(k+1)})$ [Ambainis03] via quantum walks
 $O(N^{3/4-1/\exp(k)})$ [Belovs12] via learning graphs

Lower Bounds: $\Omega(N^{2/3})$ [Aaronson-Shi01] via polynomial method

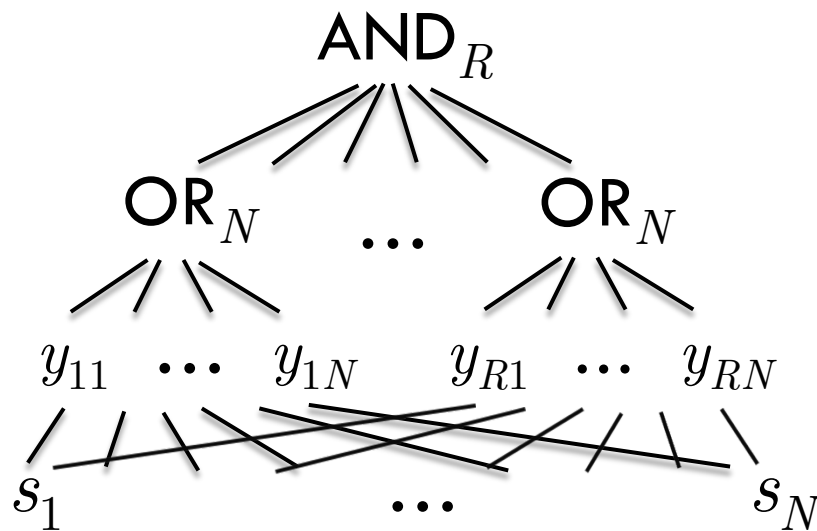
This work: $\Omega(N^{3/4-1/(2k)})$ via polynomial method

Our Results



Lower Bound Roadmap

1. Prove a hardness amplification theorem for functions in AC^0
2. Express Surjectivity, k -Distinctness, etc. as amplified versions of functions we understand



Hardness Amplification in AC^0

Theorem 1: If $\text{adeg}(f) > d$, then $\text{adeg}(F) > t^{1/2}d$
for $F = \text{OR}_t \circ f$ [B.-Thaler13, Sherstov13, BenDavid-Bouland-Garg-Kothari17]

Theorem 2: If $\text{adeg}_-(f) > d$, then $\text{adeg}_{1-2^{-t}}(F) > d$
for $F = \text{OR}_t \circ f$ [B.-Thaler14]

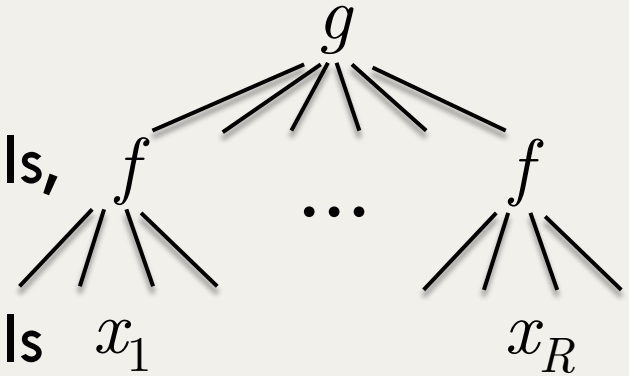
Theorem 3: If $\text{adeg}_-(f) > d$, then $\text{deg}_\pm(F) > \min\{t, d\}$
for $F = \text{OR}_t \circ f$ [Sherstov14]

Theorem 4: If $\text{adeg}_+(f) > d$, then $\text{adeg}_{1-2^{-t}}(F) > d$
for $F = \text{ODD-MAX-BIT}_t \circ f$ [Thaler14]

Theorem 5: If $\text{adeg}(f) > d$, then $\text{deg}_\pm(F) > \min\{t, d\}$
for $F = \text{APPROX-MAJ}_t \circ f$ [Bouland-Chen-Holden-Thaler-Vasudevan16]

Hardness Amplification

Theorem Template: If f is “hard” to approximate by low-degree polynomials, then $F = g \circ f$ is “even harder” to approximate by low-degree polynomials



Block Composition Barrier

Robust approximations, i.e.,

$$\text{adeg}(g \circ f) \leq O(\text{adeg}(g) \cdot \text{adeg}(f))$$

imply that block composition *cannot* give better lower bounds than \sqrt{n}

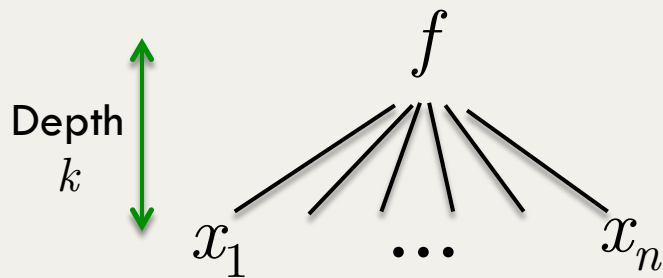
Our Work: A New Hardness Amplification Theorem for Degree

(1) An $\Omega(n^{1-\delta})$ approximate degree lower bound for AC^0

Recursive application

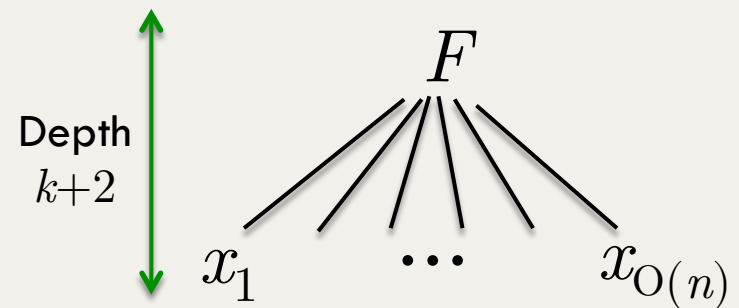
Start with:

$$\text{adeg}(f) \geq d$$



Construct:

$$\text{adeg}(F) \geq \Omega(d^{1/2} \cdot n^{1/2})$$



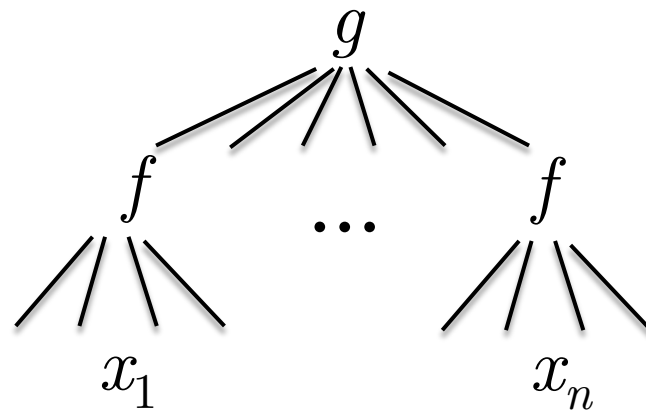
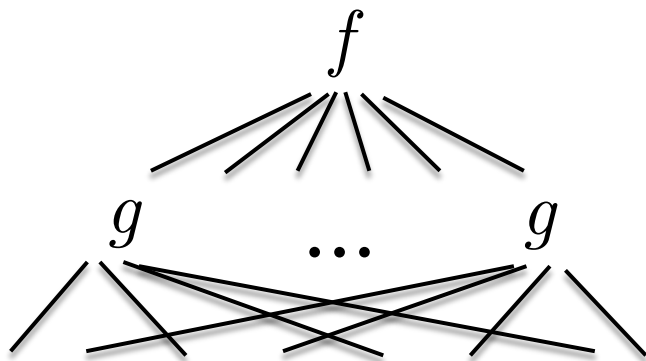
Refined & generalized application

(2) New quantum query lower bounds

Breaking the Block Composition Barrier

Prior work:

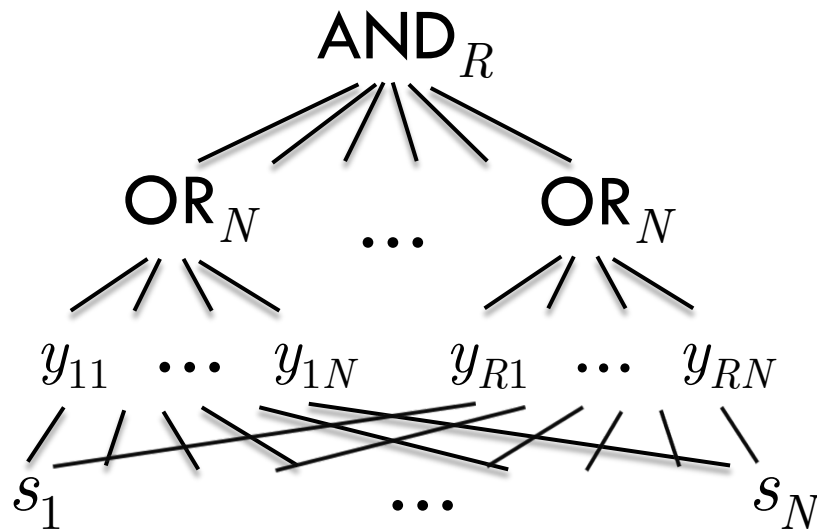
- Hardness amplification “from the top”
- Block composed functions



Our new work:

- Hardness amplification “from the bottom”
 - Non-block-composed functions

Remainder of This Talk: Lower Bound for SURJECTIVITY



Getting to Know Surjectivity

Define $\text{SURJ}_{N,R} : \{1, \dots, R\}^N \rightarrow \{0, 1\}$ by

$$\text{SURJ}_{N,R}(s_1, \dots, s_N) = 1 \quad \text{iff} \\ \text{Every } r \in [R] \text{ appears in the input list}$$

Corresponds to a Boolean function on $O(N \log_2 R)$ bits

Has quantum query complexity $\Omega(R)$ [Beame-Machmouchi10] but
approximate degree $O(R^{3/4})$ [Sherstov17]

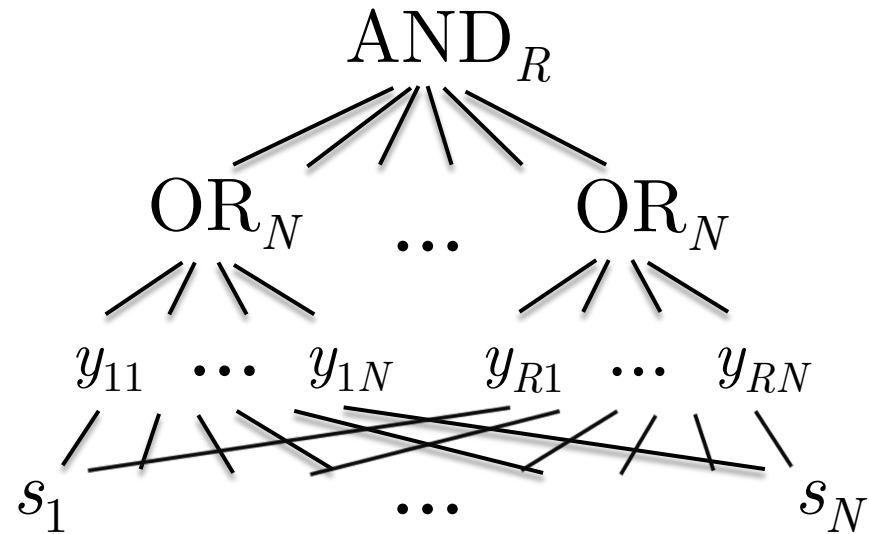
(For $N = O(R)$)

Getting to Know Surjectivity

$$\text{SURJ}_{N,R}(s_1, \dots, s_N) = 1 \quad \text{iff} \\ \text{Every } r \in [R] \text{ appears in the input list}$$

Define auxiliary variables

$$y_{r,i}(s) = \begin{cases} 1 & \text{if } s_i = r \\ 0 & \text{otherwise} \end{cases}$$



Then $\text{SURJ}_{N,R}(s_1, \dots, s_N) =$

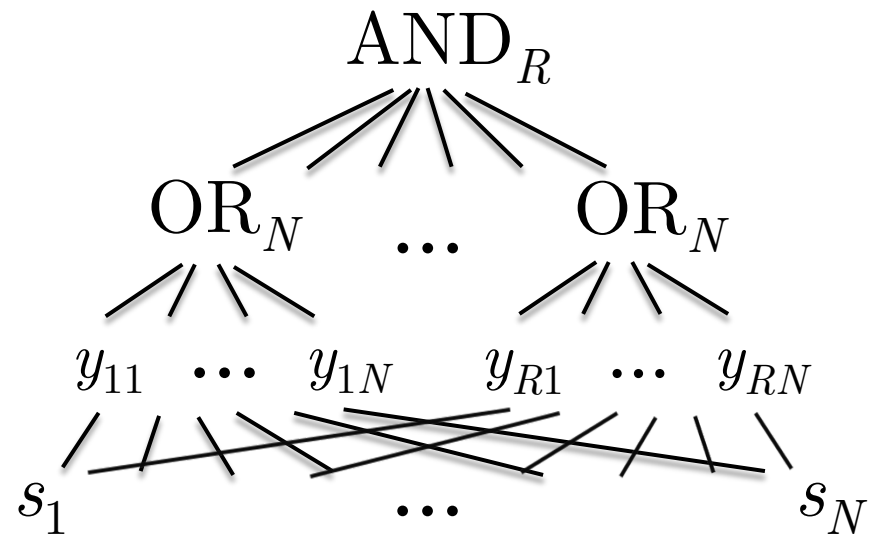
$$\text{AND}_R (\text{OR}_N (y_{11}, \dots, y_{1N}), \dots, \text{OR}_N (y_{R1}, \dots, y_{RN}))$$

Getting to Know Surjectivity

Observation: To approximate $\text{SURJ}_{N,R}$, suffices to approx. $\text{AND}_R \circ \text{OR}_N$ on inputs of Hamming weight N

Define auxiliary variables

$$y_{r,i}(s) = \begin{cases} 1 & \text{if } s_i = r \\ 0 & \text{otherwise} \end{cases}$$



Then $\text{SURJ}_{N,R}(s_1, \dots, s_N) =$

$$\text{AND}_R (\text{OR}_N (y_{11}, \dots, y_{1N}), \dots, \text{OR}_N (y_{R1}, \dots, y_{RN}))$$

Surjectivity Lower Bound

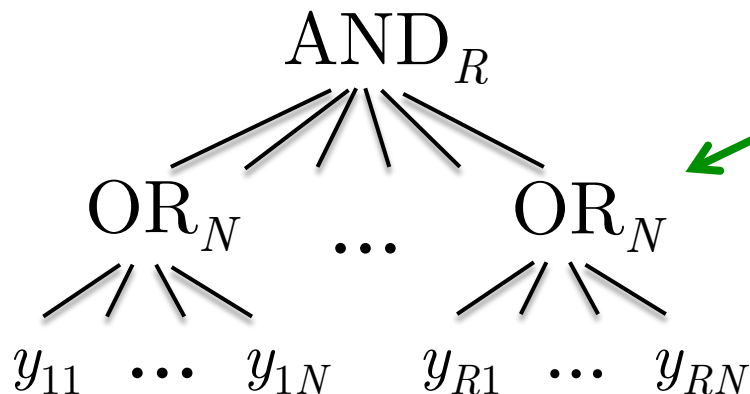
This work:

For some $N = O(R)$, $\text{adeg}(\text{SURJ}_{N,R}) = \Omega(R^{3/4})$

Stage 1: Reduce to a claim about block composed functions

Lemma: Builds on symmetrization argument of [Ambainis03]

$$\text{adeg}(\text{SURJ}_{N,R}) = \Theta(\text{adeg}(\text{AND}_R \circ \text{OR}_N^{\leq N}))$$



Promised that $|y| \leq N$

Surjectivity Lower Bound

This work:

For some $N = O(R)$, $\text{adeg}(\text{SURJ}_{N,R}) = \Omega(R^{3/4})$

Stage 2: Prove $\text{adeg}(\text{AND}_R \circ \text{OR}_N)^{\leq N} = \Omega(R^{3/4})$

Uses method of dual polynomials [Ioffe-Tikhomirov68, Sherstov07, Shi-Zhu07]

Primal

$$\begin{aligned} \min_{p, \varepsilon} \quad & \varepsilon \\ \text{s.t.} \quad & |p(x) - f(x)| \leq \varepsilon \quad \forall x \in \{0, 1\}^n \end{aligned}$$

Dual

$$\begin{aligned} \max_{\Psi} \quad & \sum_{x \in \{0, 1\}^n} (-1)^{f(x)} \Psi(x) \\ \text{s.t.} \quad & \sum_{x \in \{0, 1\}^n} |\Psi(x)| = 1 \\ & \text{deg}(p) \leq d \implies \sum_{x \in \{0, 1\}^n} p(x) \Psi(x) = 0 \end{aligned}$$

Surjectivity Lower Bound

This work:

For some $N = O(R)$, $\text{adeg}(\text{SURJ}_{N,R}) = \Omega(R^{3/4})$

Stage 2: Prove $\text{adeg}((\text{AND}_R \circ \text{OR}_N)^{\leq N}) = \Omega(R^{3/4})$

Uses method of dual polynomials [Ioffe-Tikhomirov68, Sherstov07, Shi-Zhu07]

From Justin's talk:

Can prove $\text{adeg}(\text{AND}_R \circ \text{OR}_N) = \Omega(R)$ by combining dual polynomials Ψ_{AND} and Ψ_{OR} to construct a dual polynomial $\Psi_{\text{AND-OR}}$ [B.-Thaler13, Sherstov13]

Details of Stage 2

Claim: $\text{adeg}(\text{AND}_R \circ \text{OR}_N) = \Omega(R^{3/4})$ even under the promise that $|x| \leq N$

is equivalent to

There exists a dual polynomial witnessing $\text{adeg}(\text{AND}_R \circ \text{OR}_N) = \Omega(R^{3/4})$ which is supported on inputs with $|x| \leq N$

Does the dual polynomial we already have for $\text{AND}_R \circ \text{OR}_N$ satisfy this property?

NO

Fixing the AND-OR Dual Polynomial


$$\Psi_{\text{AND-OR}}(x) = 2^R \Psi_{\text{AND}}(\text{sgn } \Psi_{\text{OR}}(x_1), \dots, \text{sgn } \Psi_{\text{OR}}(x_R)) \prod_{i=1}^R |\Psi_{\text{OR}}(x_i)|$$

Ψ_{OR} *must* be nonzero for inputs with Hamming weight up to $\Omega(N)$

$\Rightarrow \Psi_{\text{AND-OR}}$ nonzero up to Hamming weight $\Omega(RN)$

1. $\Psi_{\text{AND-OR}}$ has L_1 -norm 1 ✓
2. $\Psi_{\text{AND-OR}}$ has pure high degree $\Omega(R^{1/2}N^{1/2}) = \Omega(R)$ ✓
3. $\Psi_{\text{AND-OR}}$ has high correlation with $\text{AND}_R \circ \text{OR}_N$ ✓
4. $\Psi_{\text{AND-OR}}$ is supported on inputs with $|x| \leq N$ ✗

Fixing the AND-OR Dual Polynomial

$$\Psi_{\text{AND-OR}}(x) = 2^R \Psi_{\text{AND}}(\text{sgn } \Psi_{\text{OR}}(x_1), \dots, \text{sgn } \Psi_{\text{OR}}(x_R)) \prod_{i=1}^R |\Psi_{\text{OR}}(x_i)|$$


Ψ_{OR} *must* be nonzero for inputs with Hamming weight up to $\Omega(N)$

$\Rightarrow \Psi_{\text{AND-OR}}$ nonzero up to Hamming weight $\Omega(RN)$

Fix 1: Trade pure high degree of Ψ_{OR} for “support” size

Fix 2: Zero out high Hamming weight inputs to $\Psi_{\text{AND-OR}}$

Fix 1: Trading PHD for Support Size

For every integer $1 \leq m \leq N$, there is a dual polynomial Ψ_{OR}^m for OR_N which

- has pure high degree $\Omega(m^{1/2})$
- is supported on inputs of Hamming weight $\leq m$

$$\Psi_{\text{AND-OR}}^m(x) = 2^R \Psi_{\text{AND}}(\text{sgn } \Psi_{\text{OR}}^m(x_1), \dots, \text{sgn } \Psi_{\text{OR}}^m(x_R)) \prod_{i=1}^R |\Psi_{\text{OR}}^m(x_i)|$$

Dual polynomial $\Psi_{\text{AND-OR}}^m$

- has pure high degree $\Omega(R^{1/2} m^{1/2})$
- is supported on inputs of Hamming weight $\leq mR$

Fix 2: Zeroing Out High Hamming Weight Inputs

Dual polynomial $\Psi_{\text{AND-OR}}^m$

- has pure high degree $\Omega(R^{1/2} m^{1/2})$
- is supported on inputs of Hamming weight $\leq mR$

Suppose further that

$$\sum_{|x| > N} |\Psi_{\text{AND-OR}}^m(x)| \ll \text{negl}(R)$$

Can we post-process $\Psi_{\text{AND-OR}}^m$ to zero out inputs with Hamming weight $N < |x| \leq mR$...

...without ruining

- pure high degree of $\Psi_{\text{AND-OR}}^m$
- correlation between $\Psi_{\text{AND-OR}}^m$ and $\text{AND}_R \circ \text{OR}_N$?

YES (Follows from
[Razborov-Sherstov-08])

Fix 2: Zeroing Out High Hamming Weight Inputs

Technical Lemma (follows from [Razborov-Sherstov08])

If $0 < D < N$ and

$$\sum_{|x| > N} |\Psi_{\text{AND-OR}}^m(x)| \ll 2^{-D},$$

then there exists a “correction term” Ψ_{CORR}^m that

1. Agrees with $\Psi_{\text{AND-OR}}^m$ inputs of Hamming weight $> N$
2. Has L_1 -norm 0.01
3. Has pure high degree D

Fix 2: Zeroing Out High Hamming Weight Inputs

Claim: For $1 \leq m \leq N$,
$$\sum_{|x| > N} |\Psi_{\text{AND-OR}}^m(x)| \ll 2^{-R/m^{1/2}}$$

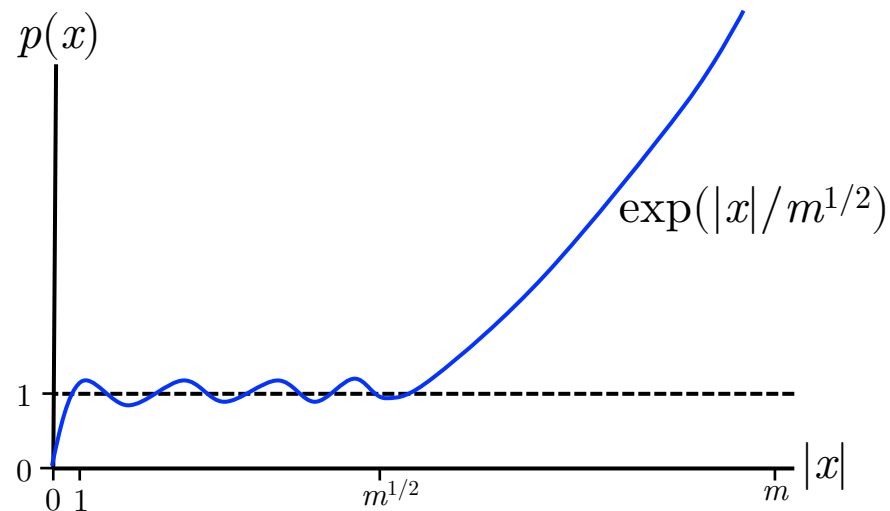
Proof idea:

Ψ_{OR}^m can be made biased toward low Hamming weight inputs:

For all $t > 0$,
$$\sum_{|x|=t} |\Psi_{\text{OR}}^m(x)| \lesssim \exp(-t/m^{1/2})$$

Primal interpretation:

Any polynomial that looks like this still has degree $\Omega(m^{1/2})$



Fix 2: Zeroing Out High Hamming Weight Inputs

Claim: For $1 \leq m \leq N$,
$$\sum_{|x| > N} |\Psi_{\text{AND-OR}}^m(x)| \ll 2^{-R/m^{1/2}}$$

Proof idea:

Ψ_{OR}^m can be made biased toward low Hamming weight inputs:

For all $t > 0$,
$$\sum_{|x|=t} |\Psi_{\text{OR}}^m(x)| \lesssim \exp(-t/m^{1/2})$$

⇒ “Worst” high Hamming weight inputs look like

$|x_1| = m^{1/2}, \dots, |x_{N/m^{1/2}}| = m^{1/2}, |x_{(N/m^{1/2})+1}| = 0, \dots, |x_R| = 0$

$$\Psi_{\text{AND-OR}}^m(x) = 2^R \Psi_{\text{AND}}(\text{sgn } \Psi_{\text{OR}}^m(x_1), \dots, \text{sgn } \Psi_{\text{OR}}^m(x_R)) \prod_{i=1}^R |\Psi_{\text{OR}}^m(x_i)|$$

Weight on such inputs looks like $2^{-N/m^{1/2}}$

Putting the Pieces Together

Dual polynomial $\Psi_{\text{AND-OR}}^m$

Fix 1

- has pure high degree $\Omega(R^{1/2} m^{1/2})$
- satisfies $\sum_{|x|>N} |\Psi_{\text{AND-OR}}^m(x)| \ll 2^{-R/m^{1/2}}$

Correction term Ψ_{corr}^m

- has pure high degree $\Omega(R/m^{1/2})$
- agrees with $\Psi_{\text{AND-OR}}^m$ inputs of Hamming weight $> N$

Balanced at $m = R^{1/2}$
 \Rightarrow PHD $\Omega(R^{3/4})$

$\Rightarrow \Psi_{\text{AND-OR}} = \Psi_{\text{AND-OR}}^m - \Psi_{\text{corr}}^m$ has

1. L_1 -norm ≈ 1
2. high correlation with $\text{AND}_R \circ \text{OR}_N$
3. pure high degree $\Omega(\min\{R^{1/2}m^{1/2}, R/m^{1/2}\})$
4. support on inputs with $|x| \leq N$

Recap of SURJECTIVITY Lower Bound

This work:

For some $N = O(R)$, $\text{adeg}(\text{SURJ}_{N,R}) = \Omega(R^{3/4})$

Stage 1: Apply *symmetrization* to reduce to

Builds on
[Ambainis03]

Claim: $\text{adeg}(\text{AND}_R \circ \text{OR}_N) = \Omega(R^{3/4})$ *even under the promise that $|x| \leq N$*

Stage 2: Prove Claim via *method of dual polynomials*

Refines AND-OR dual polynomial w/ techniques of [Razborov-Sherstov08]

Conclusions

Hardness amplification beyond block composition \Rightarrow

Nearly optimal lower bounds for AC^0

New quantum query lower bounds

Imminently forthcoming work: $\text{adeg}_\varepsilon(F) \geq \Omega(n^{1-\delta})$ for some $\varepsilon \geq 1 - \exp(-\Omega(n^{1-\delta}))$ and $F \in AC^0$

Thank you!

Open Problems:

- Approximate degree / quantum query complexity of poly-size DNF? Best lower bound: $\Omega(n^{3/4 - \delta})$
- Lower bounds for quantum problems with different structure (e.g. triangle finding, graph collision, verifying matrix products)