

# Oracle Separation of BQP and PH

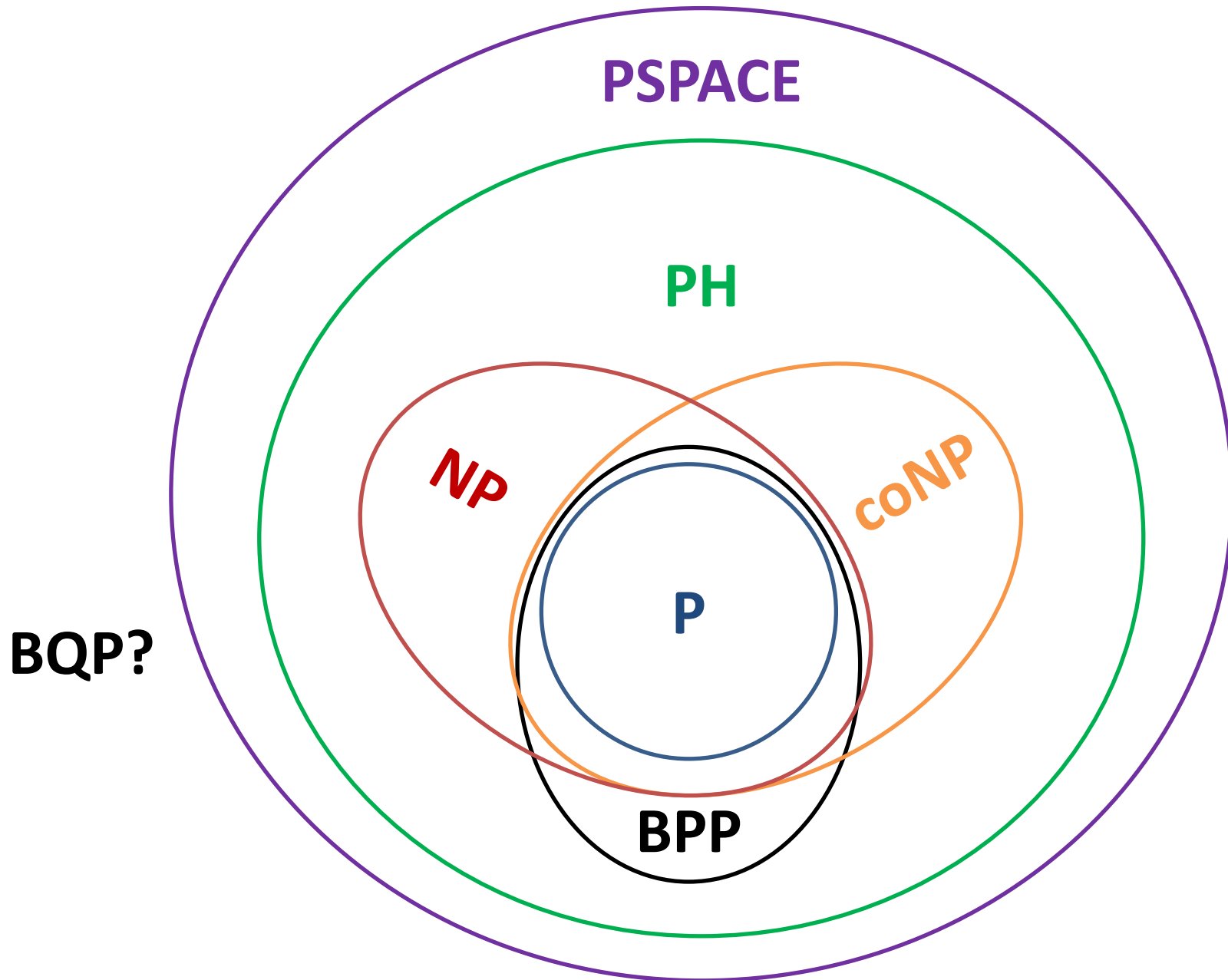
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# The Landscape of Complexity Classes



# Where does BQP fit in the landscape?

**BQP**: Bounded Error Quantum Polynomial Time

We know:  $\mathbf{BPP} \subseteq \mathbf{BQP} \subseteq \mathbf{PSPACE}$

**Oracle Separations:**

- $\exists$  oracle  $A$ :  $\mathbf{NP}^A \not\subseteq \mathbf{BQP}^A$  [BBBV'97]
- $\exists$  oracle  $A$ :  $\mathbf{BQP}^A \not\subseteq \mathbf{BPP}^A$  [BV'93]
- $\exists$  oracle  $A$ :  $\mathbf{BQP}^A \not\subseteq \mathbf{MA}^A$  [Watrous'00]

Could it be possible that  $\mathbf{BQP} \subseteq \mathbf{PH}$  ?

$\mathbf{BQP} \subseteq \mathbf{AM}$  ?

# Our Main Result: BQP vs. PH

**Recall:** a language  $L$  in **PH** iff there exists a constant  $k$ , and a poly-time computable relation  $R$  s.t.

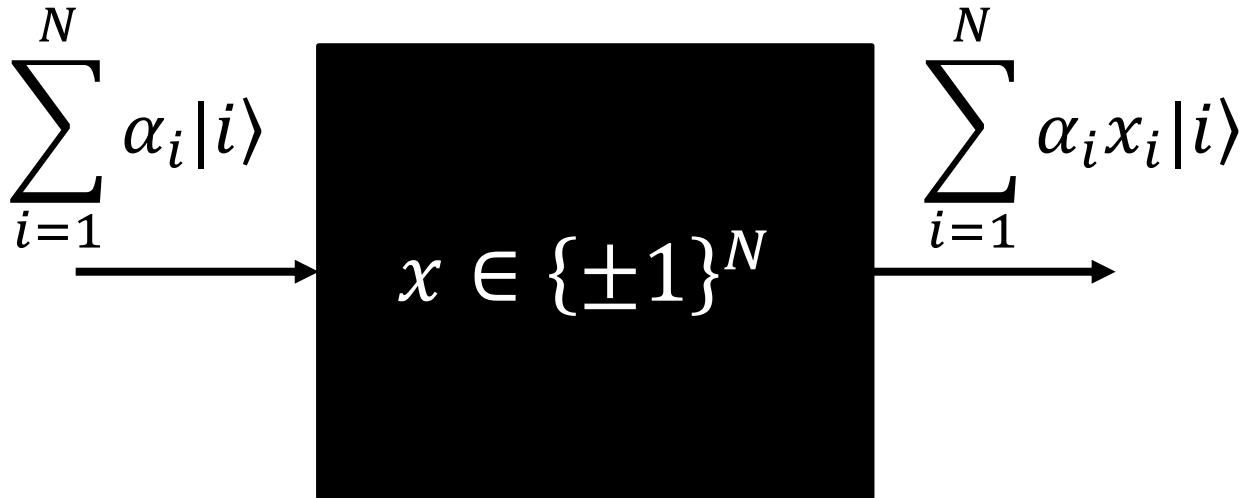
$$x \in L \iff \exists y_1 \forall y_2 \exists y_3 \dots Q_k y_k : R(x, y_1, \dots, y_k)$$

$$|y_1| + |y_2| + \dots + |y_k| \leq \text{poly}(|x|)$$

**Our Main Result:**

$\exists$  oracle  $A$ :  $\text{BQP}^A \not\subseteq \text{PH}^A$

# The Black-Box/Query Model



Complexity measure: number of queries to the black box.

Deterministic Query Complexity = **Decision Tree Complexity**

Quantum Query Complexity = Queries are made in **superposition**

**PH** analog = **AC<sup>0</sup>** circuits

**Known reductions:** Black-box separations imply oracle separations

# The Pseudorandomness Setting



**Def'n:** a distribution  $D$  is **pseudorandom** against a class of functions  $\mathcal{C}$  if

$$\forall f \in \mathcal{C}: \quad \mathbf{E}_{x \sim D}[f(x)] \approx \mathbf{E}_{x \sim U}[f(x)]$$

# The Pseudorandomness Setting



[Aaronson'10, Fefferman-Shaltiel-Umans-Viola'12]

Can you find a distribution which is pseudorandom for  $AC^0$  but not pseudorandom for **poly-log-time quantum algorithms**?

→ an oracle separation between **BQP** from **PH**



Let  $D$  be a distribution over  $\{-1,1\}^{2N}$ .

We say that an algorithm  $A$  distinguishes between  $D$  and  $U$  with **advantage**  $\alpha$  if  $\alpha = |\mathbf{E}_{x \sim D}[A(x)] - \mathbf{E}_{x \sim U}[A(x)]|$ .

**Main Result:** We present a distribution  $D$  such that:

1.  $\exists$  a **log(N)** time quantum algorithm distinguishing between  $D$  and  $U$  with advantage  $\Omega(1/\log N)$
2. Any **quasipoly(N)**-size constant-depth circuit distinguishes between  $D$  and  $U$  with advantage  $\tilde{O}\left(\frac{1}{\sqrt{N}}\right)$

Standard techniques  $\rightarrow$  amplify advantage of quantum algorithm to be **0.99** or even **1-1/poly(N)**.



# The Separating Distribution **D**

(Based on Aaronson's Forrelation distribution)

- Let  $N$  be a power of 2. Let  $\epsilon = 1/O(\log N)$ .
- Let  $G$  to be a multi-variate gaussian (**MVG**) distribution on  $\mathbb{R}^{2N}$  with zero-means and covariance matrix

$$\epsilon \cdot \begin{pmatrix} I_N & H \\ H & I_N \end{pmatrix}$$

where  $H$  is the  $N \times N$  **Hadamard** matrix with

$$H_{i,j} = \frac{1}{\sqrt{N}} \cdot (-1)^{\langle i,j \rangle}$$

**Sampling  $z' \sim D$ :**

1. Sample  $z \sim G$ , truncate each  $z_i$  to be within  $[-1,1]$
2. For  $i = 1, \dots, 2N$ , sample independently  $z'_i \in \{-1,1\}$  with  $\mathbf{E}[z'_i] = z_i$ .

# Quantum Algorithm Distinguishing **D**

[Aaronson'10, Aaronson-Ambainis'15]:

1-query  $\mathbf{O}(\log N)$ -time quantum algorithm  $Q$  s.t.

$$\Pr[Q \text{ accepts input } (x, y)] = \frac{1 + \Phi(x, y)}{2}$$

where

$$\Phi(x, y) = \frac{1}{N^{3/2}} \cdot \sum_{i=1}^N \sum_{j=1}^N (-1)^{\langle i, j \rangle} \cdot x_i \cdot y_j$$

$$\mathbf{E}_{(x, y) \sim U} [\Phi(x, y)] = 0$$

$$\mathbf{E}_{(x, y) \sim D} [\Phi(x, y)] \approx \epsilon = \Omega\left(\frac{1}{\log N}\right)$$

# $D$ is Pseudorandom for $AC^0$

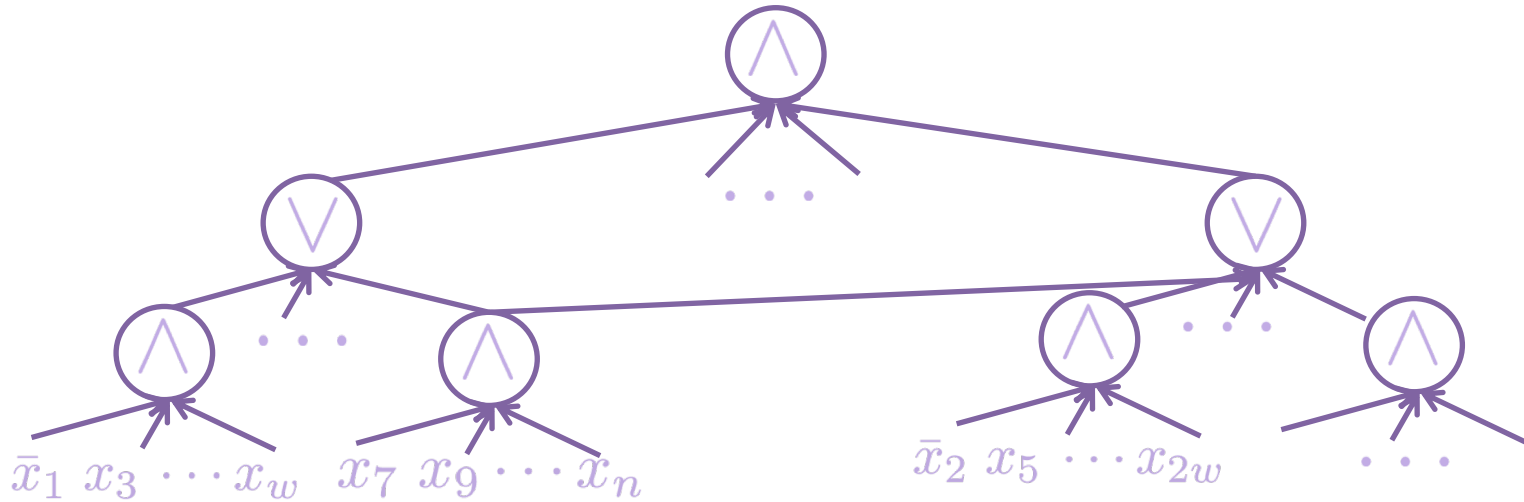
We are left to prove:

$D$  is pseudorandom for  $AC^0$ .

## Main Ingredients & Techniques:

- Fourier Analysis
- $AC^0$  circuits are well-approximated by **sparse** low-degree polynomials.
- Fractional **PRG** approach of **[CHHL18]**.
- Sum of independent Gaussians is a Gaussian.

# Bounded Depth Circuits



$AC^0[s, d]$ :

- $s$  gates (**size** of the circuit)
- depth  $d$
- alternating gates

We focus on

$$AC^0[N^{\text{polylog}(N)}, O(1)]$$

# What do we know about $AC^0$ ?

[Ajtai'83, Furst-Saxe-Sipser'84, Yao'85, Håstad '86]:

- **Parity** not in  $AC^0[N^{\text{polylog}(N)}, O(1)]$ .
  - **Parity** requires  $\exp(N^{1/(d-1)})$  size for depth  $d$ .
- $\exists$  oracle  $A$ :  $PSPACE^A \not\subseteq PH^A$

**Fourier-analytical proof technique:**

- $AC^0$  circuits can be well-approximated (in  $\ell_2$ ) by **low-degree polynomials** (over  $\mathbb{R}$ ). [Håstad '86, LMN'93]
- **Parity** cannot.

**Potential problem with the approach:**

$O(\log N)$  time quantum algorithms (**BQLogTime**) are also well-approximated by low-degree polys.

# The Difference between **BQLogTime** and **AC<sup>0</sup>**

Both **BQLogtime** & **AC<sup>0</sup>** are approximated by low-degree polynomials, but **these polynomials are different!**

**BQLogtime** can have **dense** low-degree polynomials, e.g.

$$\Phi(x, y) = \frac{1}{N^{3/2}} \cdot \sum_{i=1}^N \sum_{j=1}^N (-1)^{\langle i, j \rangle} \cdot x_i \cdot y_j$$

**[T'14]: AC<sup>0</sup>** has **sparse** low-degree approximations

$$\forall k: \sum_{S \subseteq [n], |S|=k} |\hat{f}(S)| \leq (\text{polylog } N)^k$$

# Fourier Analytical Approach – First Attempt

The Fourier expansion of  $f: \{-1,1\}^{2N} \rightarrow \{-1,1\}$ :

$$f(x) = \sum_{S \subseteq [2N]} \hat{f}(S) \cdot \prod_{i \in S} x_i$$

**Goal:**  $|\mathbf{E}_{z' \sim D}[f(z')] - \mathbf{E}_{x \sim U}[f(x)]| = \tilde{O}\left(\frac{1}{\sqrt{N}}\right)$

**Recall: Sampling  $z' \sim D$ :**

1. Sample  $z \sim G$ , truncate each  $z_i$  to be within  $[-1,1]$
2. For  $i = 1, \dots, 2N$ , sample independently  $z'_i \in \{-1,1\}$  with  $\mathbf{E}[z'_i] = z_i$

Using multilinearity of  $f$  and that whp  $\text{trunc}(z) = z$ :

$$\mathbf{E}_{z' \sim D}[f(z')] = \mathbf{E}_{z \sim G}[f(\text{trunc}(z))] \approx \mathbf{E}_{z \sim G}[f(z)]$$

→ Suffices to show  $|\mathbf{E}_{z \sim G}[f(z)] - \mathbf{E}_{x \sim U}[f(x)]| = \tilde{O}\left(\frac{1}{\sqrt{N}}\right)$

# Fourier Analytical Approach – First Attempt

$$\begin{aligned} & \mathbf{E}_{z \sim G}[f(z)] - \mathbf{E}_{x \sim U}[f(x)] \\ &= \sum_{S \subseteq [2N]} \hat{f}(S) \cdot \left( \mathbf{E}_{z \sim G} \left[ \prod_{i \in S} z_i \right] - \mathbf{E}_{x \sim U} \left[ \prod_{i \in S} x_i \right] \right) \\ &= \sum_{S \subseteq [2N], |S| \geq 1} \hat{f}(S) \cdot \mathbf{E}_{z \sim G} \left[ \prod_{i \in S} z_i \right] \\ &= \sum_{\ell=1}^N \sum_{|S|=2\ell} \hat{f}(S) \cdot \mathbf{E}_{z \sim G} \left[ \prod_{i \in S} z_i \right] \\ &\leq \sum_{\ell=1}^N \sum_{|S|=2\ell} |\hat{f}(S)| \cdot \epsilon^\ell \cdot \frac{\ell!}{\sqrt{N}^\ell} \\ &\leq \sum_{\ell=1}^N \text{polylog}(N)^{2\ell} \cdot \epsilon^\ell \cdot \frac{\ell!}{\sqrt{N}^\ell} \end{aligned}$$

Contribution of first  $\tilde{O}(\sqrt{N})$  terms:  
 $\epsilon \cdot \text{polylog}(N) / \sqrt{N}$

**Contribution of larger terms?**



# Main Technical Lemma

Suppose  $Z \sim G$  is a zero-mean **MVG** on  $\mathbb{R}^{2N}$  with

- $\forall i: \text{var}(Z_i) \leq 1/O\left(\log\left(\frac{N}{\delta}\right)\right)$
  - $\forall i, j: \text{cov}(Z_i, Z_j) \leq \delta$
- $\rightarrow$  whp  $Z \in [-1, 1]^{2N}$

Then, for any quasi-poly size constant depth **AC<sup>0</sup>** circuit  $f$ ,

$$|\mathbf{E}_{z \sim G}[f(z)] - \mathbf{E}_{x \sim U}[f(x)]| \leq \delta \cdot \text{polylog}(N)$$

Which properties of **AC<sup>0</sup>** circuits are used in the proof?

- The bound  $\sum_{|S|=2} |\hat{f}(S)| \leq \text{polylog}(N)$
- Closure under restrictions.

$G$  fools any class of functions with these two properties

# Viewing $Z \sim G$ as a result of a random walk

## A Thought Experiment:

Instead of sampling  $Z \sim G$   
at once, we sample  $t$  vectors

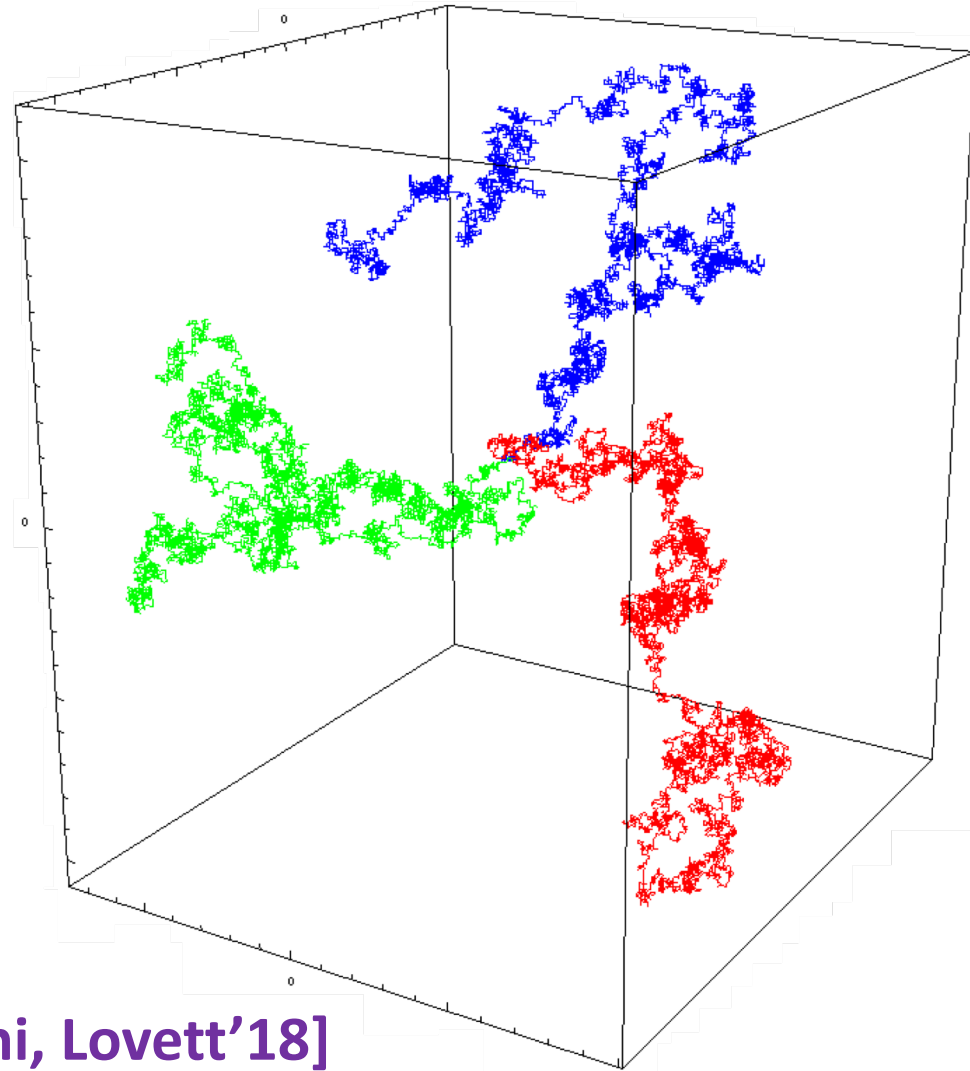
$$Z^{(1)}, \dots, Z^{(t)} \sim G$$

independently, and take

$$Z = \frac{1}{\sqrt{t}} \cdot (Z^{(1)} + \dots + Z^{(t)})$$

Based on the work of

[Chattopadhyay, Hatami, Hosseini, Lovett'18]



# Viewing $Z \sim G$ as a result of a random walk

Sample  $t$  vectors  $Z^{(1)}, \dots, Z^{(t)} \sim G$

Define  $t + 1$  hybrids:

- $H_0 = \vec{0}$
- For  $i = 1, \dots, t$

$$H_i = \frac{1}{\sqrt{t}} \cdot (Z^{(1)} + \dots + Z^{(i)})$$

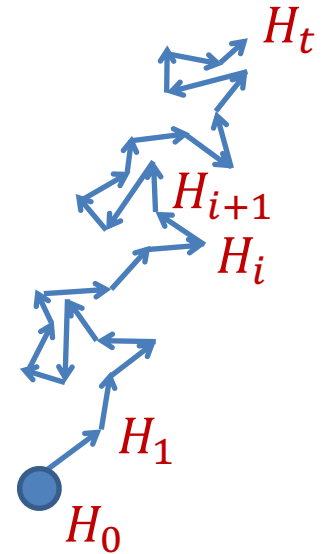
Observe:  $H_t \sim G$ .

Taking  $t \rightarrow \infty$  yields a Brownian motion.

We take  $t = \text{poly}(N)$ .

Claim: for  $i = 0, \dots, t - 1$ ,

$$|\mathbf{E}[f(H_{(i+1)})] - \mathbf{E}[f(H_i)]| \leq \frac{\delta}{t} \cdot \text{polylog}(N).$$



# Claim - Base Case

**Base Case:**

$$\begin{aligned} \mathbf{E}[f(H_1)] - \mathbf{E}[f(H_0)] &= \mathbf{E} \left[ f \left( \frac{1}{\sqrt{t}} \cdot Z^{(1)} \right) \right] - f(\vec{0}) \\ &= \sum_{\ell=1}^N \sum_{|S|=2\ell} \hat{f}(S) \cdot \mathbf{E}_{Z \sim G} \left[ \left( \frac{1}{\sqrt{t}} \right)^{2\ell} \cdot \prod_{i \in S} Z_i \right] \\ &\leq \sum_{\ell=1}^N \sum_{|S|=2\ell} |\hat{f}(S)| \cdot \frac{\delta^\ell \cdot O(\ell)^\ell}{t^\ell} \\ &\leq \frac{\delta}{t} \cdot \text{polylog}(N) + o\left(\frac{\delta}{t}\right) \end{aligned}$$

# Reducing the General Case to the Base Case

**Lemma [CHHL'18]:** for all  $z_0 \in [-1/2, 1/2]^{2N}$

$$g(z) = f(z + z_0) - f(z_0)$$

can be written as  $\mathbf{E}_\rho [f_\rho(2 \cdot z) - f_\rho(\vec{0})]$  where  $f_\rho$  is a random restriction of  $f$  (whose marginals depend on  $z_0$ ).

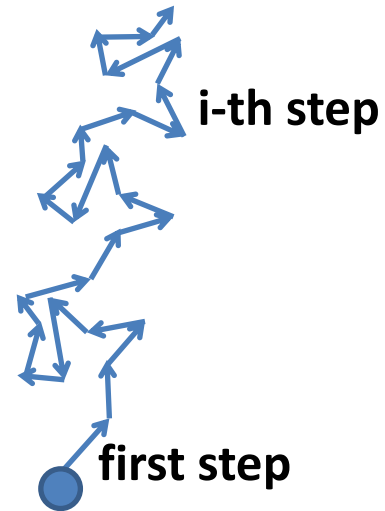
Conditioned on  $H_i \in [-1/2, 1/2]^{2N}$  (happens whp):

$$\begin{aligned} & \left| \mathbf{E}[f(H_{(i+1)})] - \mathbf{E}[f(H_i)] \right| \\ & \leq \left| \mathbf{E} \left[ f \left( H_i + \frac{1}{\sqrt{t}} Z^{(i+1)} \right) - f(H_i) \right] \right| \\ & \leq \left| \mathbf{E} \left[ f_\rho \left( \frac{2}{\sqrt{t}} \cdot Z^{(i+1)} \right) - f_\rho(\vec{0}) \right] \right| \leq \frac{4\delta}{t} \cdot \text{polylog}(N) \end{aligned}$$

# Recap: Proof by Picture

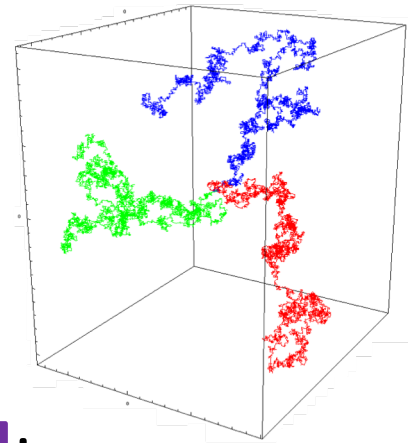
**[CHHL'18]:**  $i$ -th step  $\approx$  first step,  
using closure under restrictions.

**First Step:** Simple Fourier Analysis  
Only second level matters.



# Recap

- Defined a distribution  $D$  based on **MVG**  $G$ .
- $D$  is **not pseudorandom** for  $\log(N)$ -time **quantum** algorithms. [Aaronson'09, Aaronson-Ambainis'15]
- $D$  is **pseudorandom** for **AC<sup>0</sup>** (our contribution)  
 $|\mathbf{E}_{z \sim G}[f(z)] - \mathbf{E}_{x \sim U}[f(x)]| \leq \delta \cdot \text{polylog}(N)$ :
  - **Thought experiment**: Viewing  $Z \sim G$  as a result of a random walk with  $t$  tiny steps.
  - **AC<sup>0</sup>** circuits are well-approximated by **sparse** low-degree polynomials [T'14]
    - first step has advantage  $\left(\frac{\delta}{t}\right) \cdot \text{polylog}(N)$
  - [Chattopadhyay, Hatami, Hosseini, Lovett '18]:
    - $i$ -th step has advantage  $\left(\frac{\delta}{t}\right) \cdot \text{polylog}(N)$



# Open Problems & New results

## Follow-ups:

- [Aaronson, Fortnow]: an oracle  $A$  s.t.  
$$\text{BQP}^A \not\subseteq \text{P}^A = \text{NP}^A$$
- [Fortnow]: under our oracle  $\text{PH}$  is infinite.

## Open Problems:

- Does the original suggestion of [Aaronson'09] (without  $1/\log(N)$  noise) work?
- [Aaronson]: Find an oracle  $A$  s.t.
  - $\text{NP}^A \subseteq \text{BQP}^A$
  - $\text{PH}^A \not\subseteq \text{BQP}^A$
- [Fortnow]: Does  $\text{NP}^{\text{BQP}} \not\subseteq \text{BQP}^{\text{NP}}$ ?



# Open Problems 2: Pseudorandomness

Separate **BQLogTime** and  $\mathbf{AC}^0[\oplus]$ .

Suffices to show for all  $f$  in  $\mathbf{AC}^0[\oplus]$ :

$$\sum_{|S|=2} |\hat{f}(S)| \leq \frac{\sqrt{N}}{\text{polylog}(N)}$$

**Conjecture [CHLT'18]:** for all  $f$  in  $\mathbf{AC}^0[\oplus]$

$$\sum_{|S|=2} |\hat{f}(S)| \leq \text{polylog}(N)$$

**Claim [CHLT'18]:** **Conjecture** implies a **PRG** for  $\mathbf{AC}^0[\oplus]$  with  $\text{polylog}(N)$  seed length.

# Thank You!



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