On the complexity and verification of Random Circuit Sampling

Chinmay Nirkhe

nirkhe@cs.berkeley.edu

http://cs.berkeley.edu/~nirkhe



Joint work with



Adam Bouland UC Berkeley



Bill Fefferman
QuICS (U. Maryland/NIST)



Umesh Vazirani UC Berkeley

On the Complexity and Verification of Random Circuit Sampling A. Bouland, B. Fefferman, C. Nirkhe, U. Vazirani [Nature Physics, 2018] [arXiv:1803.04402] [ITCS 2019] [QIP 2019]

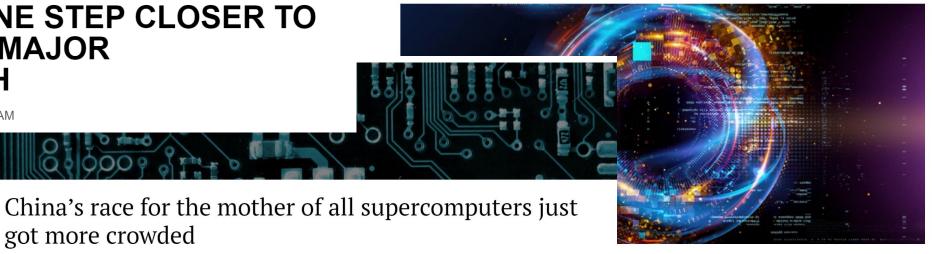
TECH & SCIENCE

REVOLUTIONARY QUANTUM **COMPUTER IS ONE STEP CLOSER TO REALITY AFTER MAJOR** BREAKTHROUGH

got more crowded

BY ARISTOS GEORGIOU ON 3/8/18 AT 9:22 AM

Is Government Ready for the Brewing **Quantum Storm?**



Baidu, Alibaba and Tencent jockey for position in the development of quantum computing, which delivers a faster and more efficient approach to processing information than today's fastest computers

Why law firms need to worry about quantum computing

BY AGNESE SMITH December 7, 2018

Safe and secure with blockchain

Will quantum computing break blockchain?

② December 12, 2018 🚨 Gary Stevens

TECH & SCIENCE

REVOLUTIONARY QUANTUM COMPUTER IS ONE STEP CLOSER TO REALITY AFTER MAJOR BREAKTHROUGH

BY ARISTOS GEORGIOU ON 3/8/18 AT 9:22 AM

Is Government Ready for the Brewing Quantum Storm?



Why is there so much hype?

Why law firms need to worry about quantum computing

BY AGNESE SMITH December 7, 2018

Safe and secure with blockchain

Will quantum computing break blockchain?

② December 12, 2018 🚨 Gary Stevens

Any "reasonable" method of computation can be efficiently simulated on a standard model (i.e. Turing machine, uniform circuits, etc.)

Any "reasonable" Quantum computation can be efficiently sixulated on a standard model (i.e. Turing machine, unito m circuits, etc.)

 $\exists \mathcal{O} \text{ s.t. } BPP^{\mathcal{O}} \neq BQP^{\mathcal{O}} \text{ [Simon}^{93}, Bernstein-Vazirani}^{93}$]

 $\exists \mathcal{O} \text{ s.t. } BPP^{\mathcal{O}} \neq BQP^{\mathcal{O}} \text{ [Simon}^{93}, Bernstein-Vazirani}^{93}$

FACTORING ∈ BQP [Shor⁹⁴]

BQP = the set of languages decidable by a polynomial time quantum algorithm

 $\exists O \text{ s.t. } \text{RPP}^{O} \neq \text{ROP}^{O} \text{ [Simon}^{93} \text{ Bernstein-Vazirani}^{93}]$

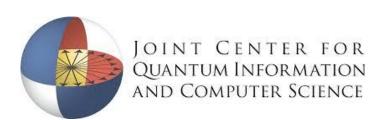
So there is theoretical evidence, but is there anything tangible?

вор — тпе ser or ranguages aectaable by a polynomial time quantum algorithm

rigetti

Experimental progress





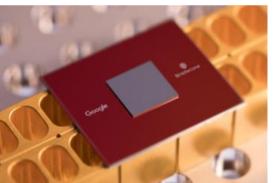


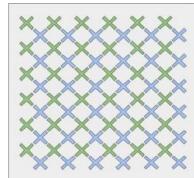












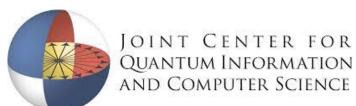


72-qubit Bristlecone chip

Experimental progress

















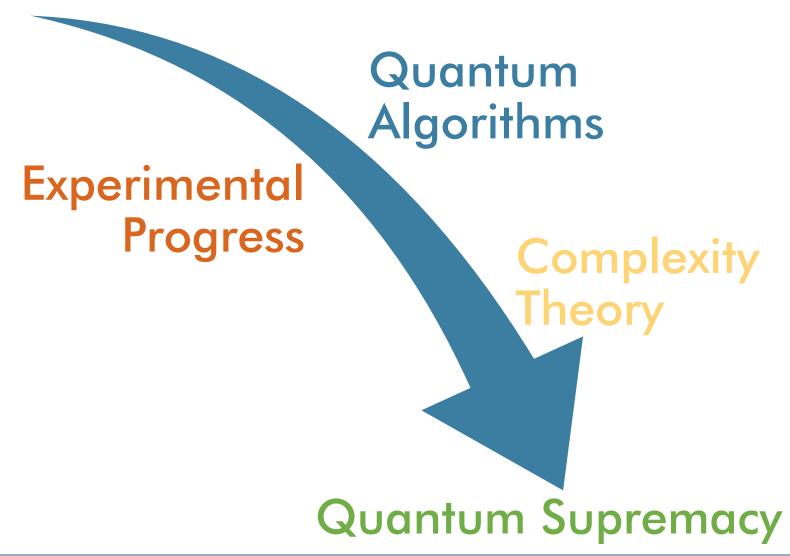




72-qubit Bristlecone chip

IBM Q

Oracle Separation



Quantum supremacy proposal

A practical demonstration of a quantum computation which is

- 1. Experimentally feasible
- 2. Has theoretical evidence of hardness
- 3. Verifiable

Quantum supremacy proposal

A practical demonstration of a quantum computation which is

NEWS IN BRIEF QUANTUM PHYSICS

Google moves toward quantum supremacy with 72-qubit computer

IBM and Intel recently debuted similarly sized chips



Experimentally feasible

2

"an experimental violation of the extended Church-Turing thesis" – U. Vazirani

Why factoring is not the right proposal

The speedups come from carefully engineered interference patterns with large amounts of constructive and destructive interference

This behavior is far from "typical".
It's hard to make this happen in the lab!!

Which is hard to generate on the currently available noisy intermediate scale quantum devices

"Proving a quantum system's computational power by having it factor integers is a bit like proving a dolphin's intelligence by teaching it to solve arithmetic problems"
[Aaronson-Arkhipov¹¹]

Complexity theory inspired supremacy proposals

Problems for which no efficient classical algorithms exist (perhaps under complexity-theoretic conjectures)

Example: Boson Sampling [Aaronson¹¹]

Proves efficient classical algorithms cannot exist unless PH-collapses

Experimentally inspired supremacy proposals

Problems which we can experimentally test in the near future (~10 years)

Example: Random Circuit Sampling [BIS+16]

Near-term experimentally feasible due to high-quality superconducting qubits

A Quantum Supremacy Proposal

Random Circuit Sampling

Given the description of a quantum circuit C, sample from the output distribution of the quantum circuit.

Part 0: What is quantum computing?

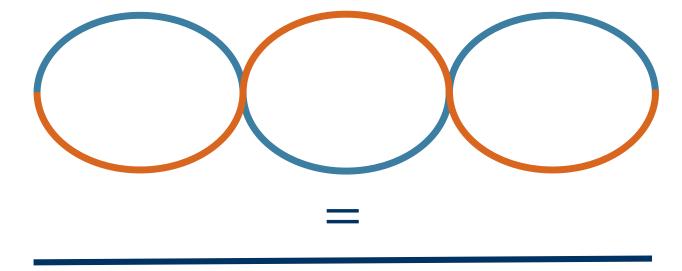
It's computing (really, information processing) based on the principles of quantum mechanics rather than classical physics.

Quantum mechanics is a description of nature

- Formulated to explain the behavior of subatomic particles.
- QM has been spectacularly successful in explaining microscopic physical phenomena.

Quantum computers run in *superposition*. It's like running probabilistically except there can be negative (complex) probabilities.

Remember physics?



The state of a deterministic computation is a binary string

$$x \in \{0,1\}^m$$

The state of a randomized computation is a probability distribution

$$\{p_x\}_{x\in\{0,1\}^m}$$
 $\sum_x p_x = 1; p_x \ge 0$

The state of a quantum computation is a superposition

$$\sum_{x} \alpha_{x} |x\rangle \qquad \sum_{x} \alpha_{x}^{2} = 1; \ \alpha_{x} \in \mathbb{C}$$

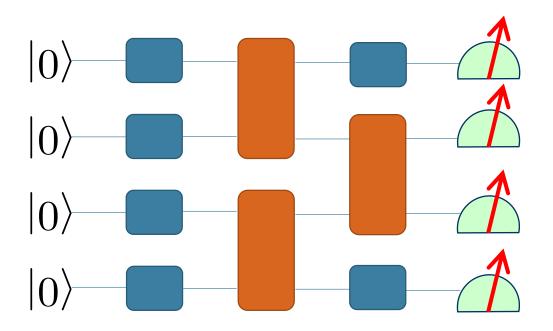
Quantum computers are realized by measurement. Classical numbers which we can read.

At a high level, quantum computing gives us some of the power of parallel computation without multiple processors.

Some problems have good quantum algorithms, while we believe that for some quantum offers no improvement.

Part 1: Classical hardness of Random Circuit Sampling

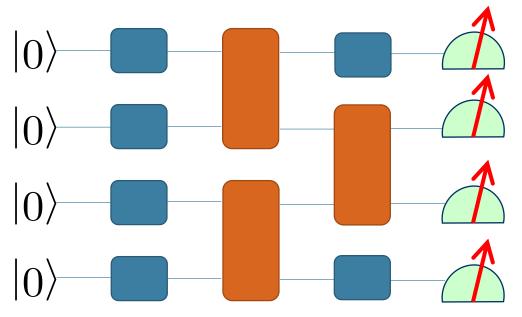
Fix an architecture over quantum circuits



Given a circuit from the architecture, sample from its output probability distribution

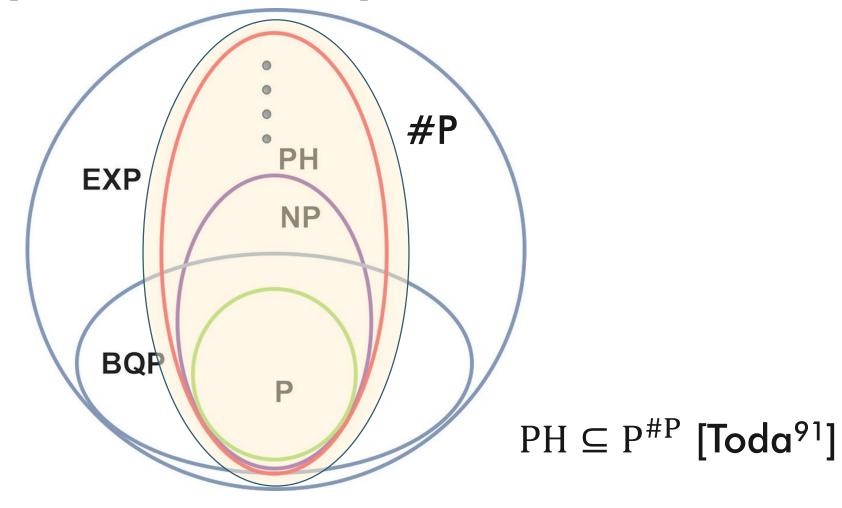
24

Sampling from the exact output distribution of a quantum circuit is #P-hard



Trick: Since proving #P-hardness, by Toda's Theorem can use PH reductions instead of just P reductions

Recall the polynomial hierarchy...





Exact classical sampling from quantum circuits would give us:

$$P^{\#P} \subseteq BPP^{NP}$$

Contradicts the non-collapse of the PH:

$$\mathsf{BPP}^{\mathsf{NP}} \subseteq \Sigma_3 \subsetneq \mathsf{PH} \subseteq \mathsf{P}^{\#\mathsf{P}}$$

Toda's Theorem

Pf: Estimating output probabilities is #P-hard. Apply BPP^{NP} reduction due to Stockmeyer's Thm⁸⁵ to get sampling is also #P-hard.

Therefore, exact quantum sampling is #P-hard under BPPNP-reductions

But...

No quantum device would exactly sample from the output distribution due to noise!

So in order to make this argument, we have to show that no classical sampler can even approximately sample from the same distribution!

By construction, the "hardness" of our circuit was in Pr(0).

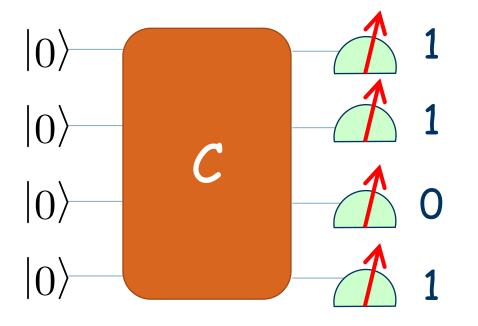
If an adversary knew that, they could generate an approximate sampler that never outputted 0.

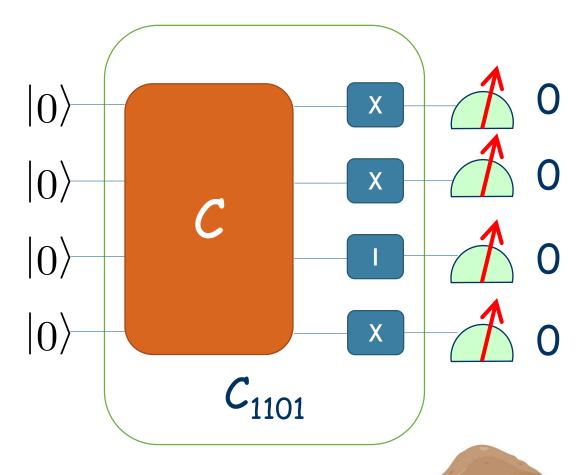
Thereby, short-circuiting the hardness proof.

We want to embed the hardness across all the outputs of the probability distribution.

We want a robustness condition: Being able to compute most probabilities should be #P-hard.

Hiding





 $Pr[C \text{ outputs } 1101] = Pr[C_{1101} \text{ outputs } 0000]$

We want a robustness condition: Being able to compute most probabilities should be #P-hard.

We want a robustness condition: Being able to compute $\Pr_0(C)$ for most circuits C should be #P-hard.

$$Pr_0(C) = prob. C outputs 0$$

What known problem has such a property?

$$\operatorname{perm}(M) = \sum_{\sigma \in S_n} \prod_{j=1}^n M_{j,\sigma(j)}$$

Theorem [Lipton⁹¹,GLR+⁹¹]: The following is #P-hard: For sufficiently large q, given uniformly random $n \times n$ matrix M over \mathbb{F}_q , output $\operatorname{perm}(M)$ with $\operatorname{probability} > \frac{3}{4} + \frac{1}{\operatorname{poly}(n)}$

Permanent is avg-case hard $\operatorname{perm}(A) = \sum_{\sigma \in S_n} \prod_{j=1} A_{j,\sigma(j)}$

perm(A) is a degree n polynomial in the matrix entries of A

Choose R a random matrix. Let M(t) = A + Rt. M(0) = A and M(t) for $t \neq 0$ is uniformly random. perm(M(t)) is a degree n polynomial in t

Choose random $t_1, ..., t_{n+1}$, calculate $perm(M(t_i))$ Interpolate the polynomial perm(M(t)). Output perm(M(0))

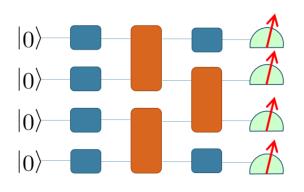
Permanent is avg-case hard $\operatorname{perm}(A) = \sum_{\sigma \in S_n} \prod_{j=1} A_{j,\sigma(j)}$

Assume we can calculate perm(R) for random R with probability > 1 - 1/(3n + 3).

By union bound, we calculate perm(A) with probability 2/3. Since permanent is worst-case #P-hard [Valiant⁷⁹], This proves statement for probability 1 - 1/(3n + 3).

Better interpolation techniques bring the probability down to $\frac{3}{4} + \frac{1}{\text{poly}(n)}$.

Goal: find a similar polynomial structure in the problem of Random Circuit Sampling



High-level idea

Feynman Path Integral:

$$\langle y_m | C | y_0 \rangle = \sum_{y_1, y_2, \dots, y_{m-1} \in \{0,1\}^n} \prod_{j=1}^{n} \langle y_j | C_j | y_{j-1} \rangle.$$

Quantum analog of space-efficient brute-force evaluation of a circuit

m

Feynman Path Integral

$$\langle 0|C|0\rangle = \sum_{y_1,y_2,...,y_{m-1}\in\{0,1\}^n} \prod_{j=1}^{n} \langle y_j|C_j|y_{j-1}\rangle.$$

m

Then $Pr_0(C)$ is a low-degree polynomial in the gate entries. We want to apply a similar interpolation technique as permanents.

$$\langle 0|C|0\rangle = \sum_{y_1,y_2,...,y_{m-1}\in\{0,1\}^n} \prod_{j=1}^{n} \langle y_j|C_j|y_{j-1}\rangle.$$

Consider the circuit C(t) formed by changing each gate C_i to $C_i + tH_i$ for random gate H_i .

Just like permanent!

But, $C_i + tH_i$ isn't a quantum gate!

$$\langle 0|C|0\rangle = \sum_{y_1,y_2,...,y_{m-1}\in\{0,1\}^n} \prod_{j=1} \langle y_j|C_j|y_{j-1}\rangle.$$

Consider the circuit $C(\theta)$ formed by changing each gate

$$C_i \mapsto C_i H_i e^{-i\theta h_i}$$
 where $h_i = -i \log H_i$

- 1. C(1) = C
- 2. For small θ , circuit looks θ -close to random!
- 3. Not a low-degree polynomial in θ

$$\langle 0|C|0\rangle = \sum_{y_1,y_2,...,y_{m-1}\in\{0,1\}^n} \prod_{j=1}^{n} \langle y_j|C_j|y_{j-1}\rangle.$$

Consider the circuit C(t) formed by changing each gate C_i to $C_i + tH_i$ for random gate H_i .

Just like permanent!

But, $C_i + tH_i$ isn't a quantum gate!

$$\langle 0|C|0\rangle = \sum_{y_1,y_2,...,y_{m-1}\in\{0,1\}^n} \prod_{j=1}^n \langle y_j|C_j|y_{j-1}\rangle.$$

Taylor Series!

Replace
$$e^{-i\theta h_i}$$
 with
$$\sum_{k=0}^{\mathsf{poly}(n)} \frac{(-i\theta h_i)^k}{k!}$$

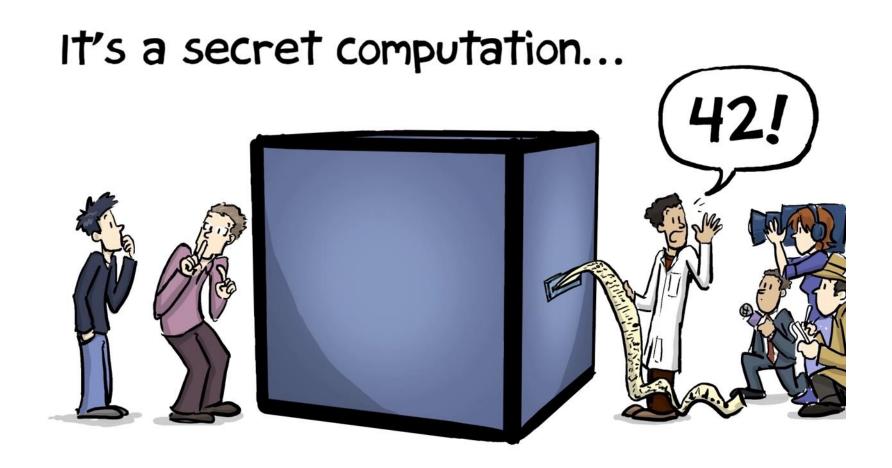
- 1. $C(1) \approx C$
- 2. For small θ , circuit looks θ -close to random!
- 3. A low-degree polynomial in θ
- 4. For more complicated technical reasons, this is a necessary, but not sufficient, proof of average-case hardness.

Theorem: On average, it is hard to exactly sample from the output of random circuits.

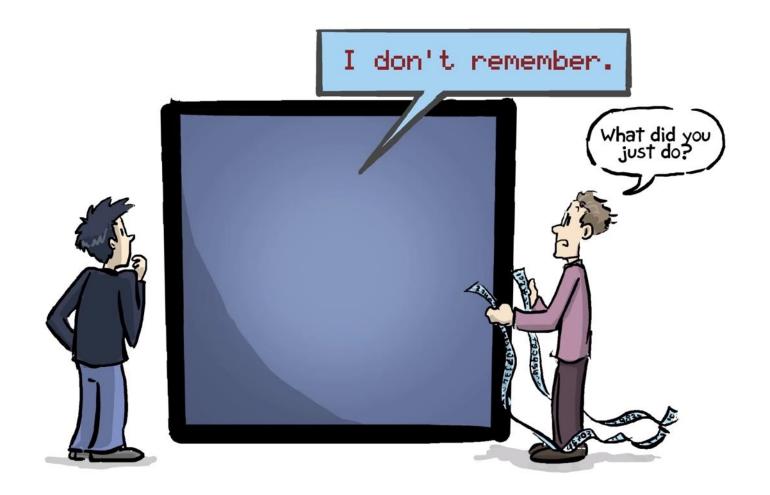
This puts Random Circuit Sampling on par with the best known supremacy proposals.

Part 2: Verification of Random Circuit Sampling

How do we know if our quantum computer is doing what it says its doing?



phdcomics.com & Caltech IQIM



phdcomics.com & Caltech IQIM

Sweet spot in verification

Toolbox

Noisy 50 qubit quantum computer

Super-computers capable of 2⁶⁰ size computations

Compromise: OK with exponential post processing time on supercomputer to compute "a few" ideal output probabilities for "intermediate" size quantum computers (n = 50 qubits)

Constraint: can only take a small (poly(n)) number of samples from the quantum device

Challenge: Complexity arguments require closeness in total variation distance. But we can't hope to unconditionally verify this with few samples from the device.

Candidate test: Cross-Entropy [Boixo + 16]

$$CE(p_{\text{dev}}, p_{\text{id}}) = \sum_{x} p_{\text{dev}}(x) \log \frac{1}{p_{\text{id}}(x)} = \mathbb{E}_{p_{\text{dev}}} \log \frac{1}{p_{\text{id}}}$$

Can be approximated in a few samples

- Sample $x_1, ..., x_k$ from quantum device
- Use exponential time to calculate $p_{id}(x_i)$
- Estimate CE
- If score is close enough to expected CE_{ideal}, then accept.

Candidate test: Cross-Entropy [Boixo + 16]

This is a one-dimensional projection of high-dimensional information

Does not verify closeness in total-variation distance

Theorem: If cross-entropy is close to ideal and $H(p_{\text{dev}}) \geq H(p_{\text{id}})$ then the output distribution is close to ideal in total variation distance

Candidate test: Binned output generation

