

# Characterizing Quantum Supremacy in Near-Term Devices

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ICTP

August 27th

# Quantum Supremacy

J. Preskill, 2012

With a quantum device

- perform a well-defined computational task
- beyond the capabilities of state-of-the-art classical supercomputers
- in the near-term
  - without error correction.

Not necessarily solving a practical problem.

# Approaches to quantum supremacy

- Optimization of a classical function:
  - Quantum Annealing.
  - Quantum Approximate Optimization Algorithm (E. Farhi et. al.).
- Non-simulable Hamiltonian Evolution.
- Variational Quantum Eigensolver (Ground state energy of a Hamiltonian).
- **Approximate sampling from a well defined distribution:**
  - Commuting Quantum Circuits (M. Bremner et. al.).
  - BosonSampling. (Aaronson and Arkhipov).
  - **Random Universal Circuits.** “Randomized benchmarking for complex circuits.”

# Requisites for quantum supremacy in the near-term

- Classically, nothing must work for the computational task, except direct simulation of quantum evolution.
  - Cost exponential in size of Hilbert space.
  - Typical of **chaotic systems**.
- Specific figure of merit for the computational task.
  - We should measure the figure of merit up to quantum supremacy frontier.
  - Naturally related to fidelity.
- Well understood extrapolation of figure of merit beyond the quantum supremacy frontier where it can not be measured. (Unfortunately, we lack witness.)
- Predictions from theory for figure of merit.
- Relation to Computational Complexity is a plus.
  - Formal Computational Complexity is asymptotic, requires error correction (Strong Church-Turing Thesis).

# Random Universal Quantum Circuits

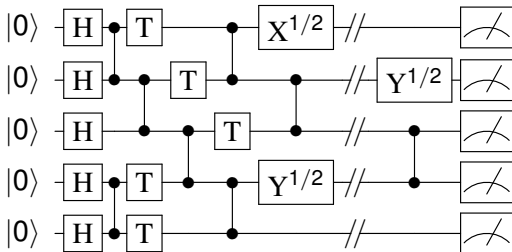


Figure: Vertical lines correspond to controlled-phase gates .

- Random quantum circuits are examples of **quantum chaos**.
- Classically sampling  $p_U(x) = |\langle x | U | 0 \rangle|^2$  requires direct simulations. Cost in  $2D$  exponential in  $\propto \min(n, d\sqrt{n})$ , depth  $d$ , qubits  $n$ . (With  $7 \times 7$  qubits requires  $d \simeq 25$ .)
- Good **benchmark for quantum computers**.
- New results in computational complexity.

# Porter-Thomas distribution

- (Pseudo-)random circuit  $U$

$$|\Psi\rangle = U|0\rangle = \sum_{j=1}^N c_j |x_j\rangle .$$

- Sample the output distribution with probabilities

$$p_i = |c_i|^2 = |\langle x_i | U | \Psi \rangle|^2 .$$

- Real and imaginary parts of  $c_i$  are distributed (quasi) uniformly on a  $2N$  dimensional sphere (Hilbert space) if the circuit (or Hamiltonian evolution) has sufficient depth (evolution time).
  - The distribution of  $c_i$  is, up to finite moments, Gaussian with mean 0 and variance  $\propto 1/N$  (random unitary matrices, delocalization, level repulsion...).
- Porter-Thomas distribution:  $\Pr(Np) = e^{-Np}$ .

# Porter-Thomas distribution

Histogram of the output distribution for different values of the two-qubit gate error rate  $r$ .

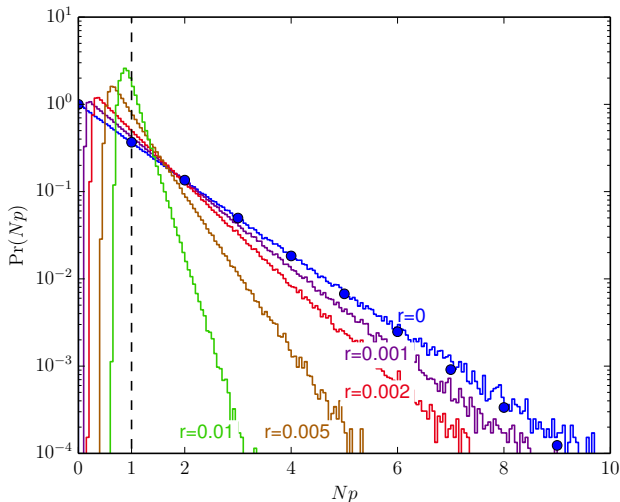


Figure: Circuit with  $5 \times 4$  qubits (2D lattice) and depth 25.

# Verification and uniformity test

- The PT distribution is very flat:  $p(x_j) \sim 1/N$ .
- The  $\ell_1$  distance between PT and uniform distribution is

$$\sum_j |p(x_j) - 1/N| = 2/e .$$

- If we calculate  $p(x_j)$  given circuit  $U$ , we can distinguish these distributions with a constant number of measurements.
- If we don't know anything about  $p(x_j)$  (black-box setting) we need  $\Theta(\sqrt{N})$  measurements.
- There is no polynomial witness for this sampling problem. This problem is much harder than NP.
  - This is required for near-term (few qubits) supremacy.



# Sampling from ideal circuit $U$

Sample  $S = \{x_1, \dots, x_m\}$  of bit-strings  $x_j$  from circuit  $U$  (measurements in the computational basis).

$$\log \Pr_U(S) = \sum_{x_j \in S} \log p_U(x_j) = -m H(p_U) + O(m^{1/2}),$$

where  $H(p_U)$  is the entropy of PT

$$H(p_U) = - \int_0^{\infty} p N^2 e^{-Np} \log p \, dp = \log N - 1 + \gamma.$$

and  $\gamma \simeq 0.577$ .

# Sampling with polynomial classical circuit $A_{\text{pcl}}(U)$

A *polynomial* classical algorithm  $A_{\text{pcl}}(U)$  produces sample  $S_{\text{pcl}} = \{x_1^{\text{pcl}}, \dots, x_m^{\text{pcl}}\}$ . The probability  $\Pr_U(S_{\text{pcl}})$  that this sample  $S_{\text{pcl}}$  is observed from the output  $|\psi\rangle$  of the circuit  $U$  is

$$\log \Pr_U(S_{\text{pcl}}) = -m H(\rho_{\text{pcl}}, \rho_U) + O(m^{1/2}),$$

where

$$H(\rho_{\text{pcl}}, \rho_U) \equiv - \sum_{j=1}^N \rho_{\text{pcl}}(x_j|U) \log \rho_U(x_j)$$

is the cross entropy.

# Sensitivity to single Pauli error

A single Pauli error (almost) destroys the output distribution  $p_U$ .

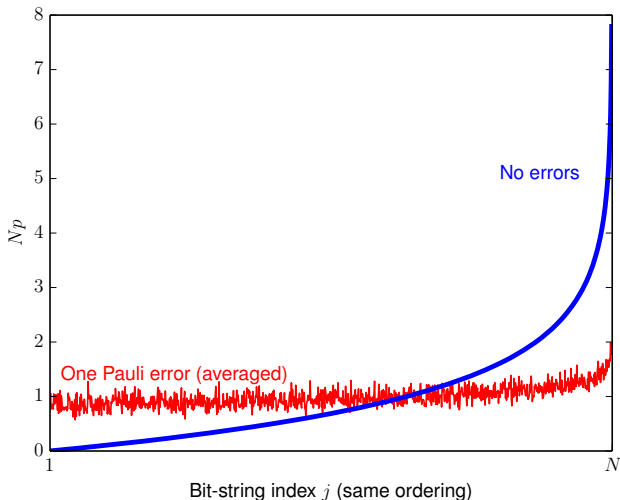


Figure: Blue line shows sorted probabilities  $p_U(x_j)$  (universal quantum chaos distribution, Porter-Thomas). Red line average of a single Pauli error in all different locations, same ordering.

## Sampling with polynomial classical circuit $A_{\text{pcl}}(U)$ (II)

We are interested in the average over  $\{U\}$  of random circuits (or chaotic evolutions)

$$\mathbb{E}_U [\mathbf{H}(\rho_{\text{pcl}}, \rho_U)] = \mathbb{E}_U \left[ \sum_{j=1}^N \rho_{\text{pcl}}(x_j|U) \log \frac{1}{\rho_U(x_j)} \right] .$$

Because  $U$  is chaotic, Hilbert space has exponential dimension, and  $A_{\text{pcl}}(U)$  is polynomial, we conjecture that  $\rho_{\text{pcl}}$  and  $\rho_U$  are (almost) **uncorrelated** (more reasons later). We can take averages independently.

$$-\mathbb{E}_U [\log \rho_U(x_j)] \approx - \int_0^\infty N e^{-Np} \log p \, dp = \log N + \gamma .$$

$$\mathbb{E}_U [\mathbf{H}(\rho_{\text{pcl}}, \rho_U)] = \log N + \gamma \equiv \mathbf{H}_0 .$$

# Cross entropy difference

- The average cross entropy of a polynomial classical algorithm is the same as for a uniform distribution  $p(x) = 1/N$ .
- For algorithm  $A$  (quantum or classical of any cost) define the **cross entropy difference**

$$\begin{aligned}\alpha &\equiv \Delta H(p_A) \equiv \log N + \gamma - H(p_A, p_U) \\ &= \sum_j \left( \frac{1}{N} - p_A(x_j|U) \right) \log \frac{1}{p_U(x_j)} .\end{aligned}$$

- The cross entropy goes between  $\alpha = 0$  for no correlation, and  $\alpha = 1$  for the ideal circuit.

# Cross entropy and fidelity

- The output of an evolution with fidelity  $\tilde{\alpha}$  is

$$\rho = \tilde{\alpha} U |0\rangle\langle 0| U^\dagger + (1 - \tilde{\alpha}) \sigma_U,$$

with  $p_{\text{exp}}(x) = \langle x | \rho | x \rangle = \tilde{\alpha} p_U(x) + (1 - \tilde{\alpha}) \langle x | \sigma_U | x \rangle$ .

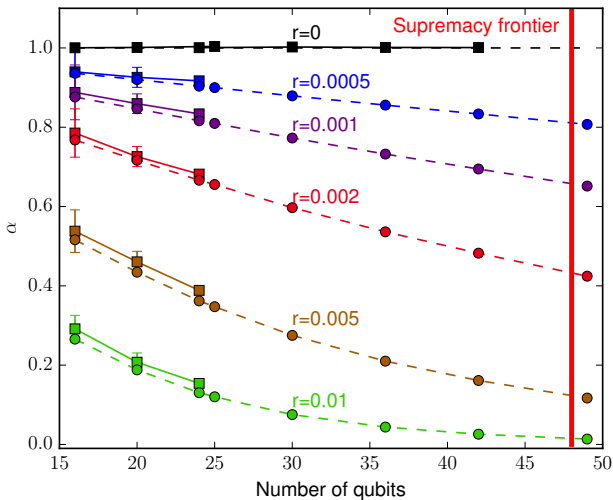
- We again conjecture that  $\langle x | \sigma_U | x \rangle$  is **uncorrelated** with  $p_U(x)$ .

$$\begin{aligned} \alpha &= \mathbb{E}_U[\Delta H(p_{\text{exp}})] \\ &= H_0 + \sum_j (\tilde{\alpha} p_U(x_j) + (1 - \tilde{\alpha}) \langle x_j | \sigma_U | x_j \rangle) \log p_U(x_j) \\ &= H_0 - \tilde{\alpha} H(p_U) - (1 - \tilde{\alpha}) H_0 = \tilde{\alpha}. \end{aligned}$$

- The cross entropy  $\alpha$  approximates the fidelity  $\tilde{\alpha}$ .

# Numerics and theory for realistic 2D circuits

Cross entropy difference  $\square$  and estimated fidelity  $\circ$ .



$r$  is two-qubit gate error rate.  $\alpha = 1$  for chaotic state.  $d = 25$ .

# Experimental proposal

- 1 Implement a random universal circuit  $U$  (chaotic evolution).
- 2 Take large sample  $\mathcal{S}_{\text{exp}} = \{x_1^{\text{exp}}, \dots, x_m^{\text{exp}}\}$  of bit-strings  $x$  in the computational basis ( $m \sim 10^3 - 10^6$ ).
- 3 Compute quantities  $\log p_U(x_j^{\text{exp}})$  with supercomputer.

Cross entropy difference (figure of merit)

$$\alpha = \frac{1}{m} \sum_{j=1}^m \log p_U(x_j^{\text{exp}}) + \log 2^n + \gamma \pm \frac{\kappa}{\sqrt{m}}, \quad \kappa \simeq 1, \gamma = 0.577$$

Measure and **extrapolate**  $\alpha$  (size, depth,  $T$  gates).

Fit to theory:  **$\alpha$  approx. circuit fidelity**, chaotic state very sensitive to errors.

$$\alpha \approx \exp(-r_1 g_1 - r_2 g_2 - r_{\text{init}} n - r_{\text{mes}} n),$$

$r_1, r_2 \ll 1$  one and two-qubit gates Pauli error rates,  $g_1, g_2 \gg 1$  number of one and two-qubit gates,  $r_{\text{init}}, r_{\text{mes}} \ll 1$  initialization and measurement error rates.



# Convergence to chaos

Depth required for PT distribution, in 2D is  $\propto \sqrt{n}$ .

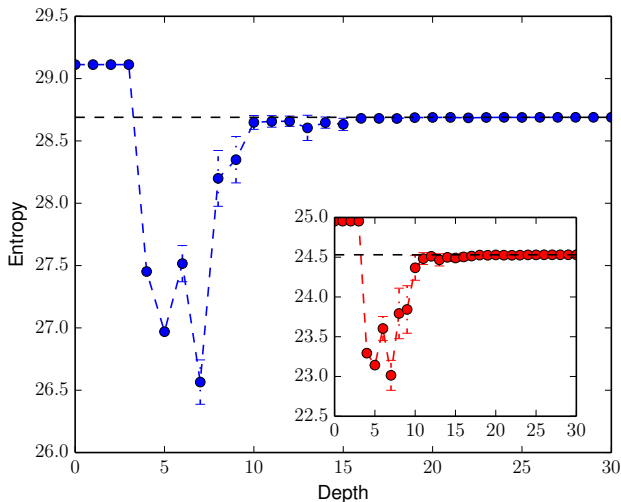


Figure: 2D circuit  $7 \times 6$  qubits. Inset  $6 \times 6$  qubits.

Dashed line is known  $H(p_U)$  for PT.

# Convergence to chaos (II)

Moments of  $p_U$  converge to PT distribution.

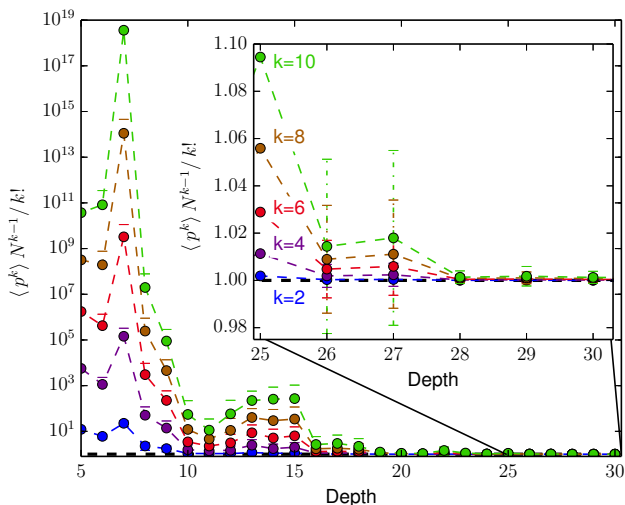


Figure: Moments  $\langle p^k \rangle$  with  $k = 2, 4, 6, 8, 10$ , normalized to 1 for PT distribution.  $7 \times 6$  circuit.

# Convergence to chaos (III)

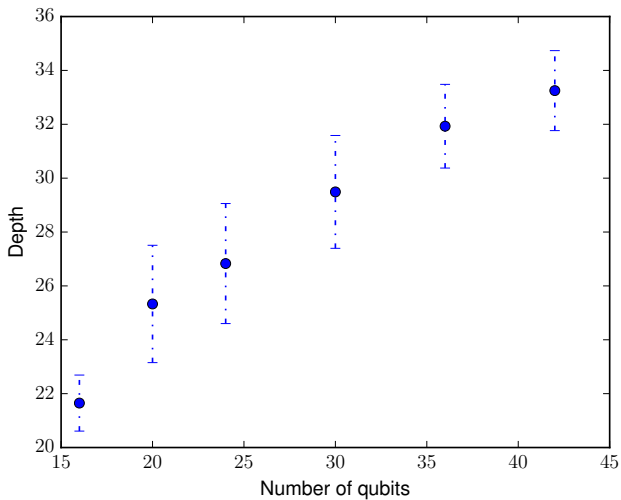


Figure: First cycle such that the entropy remains within  $2^{-n/2}$  of PT entropy.

# Complex Ising models from universal circuits

- As in a path integral, the output amplitude of  $U$  is

$$\langle x | U | 0 \rangle = \sum_{\{s^t\}} \prod_{t=0}^d \langle s^t | U^{(t)} | s^{t-1} \rangle, \quad |s^d\rangle = |x\rangle .$$

where  $|s^t\rangle = \otimes_{j=1}^n |s_j^t\rangle$  is the computational basis,  $s_j^t = \pm 1$ , and  $U^{(t)}$  are gates at clock cycle  $t$ .

- Gates give Ising couplings between spins  $s_j^k$ , like in path integral QMC. For instance, for  $X^{1/2}$  gates

$$\frac{i\pi}{4} H_s^{X^{1/2}}(x) = \frac{i\pi}{2} \sum_{j=1}^n \sum_{k=0}^{d(j)} \alpha_j^k \frac{1 + s_j^{k-1} s_j^k}{2} .$$

where  $\alpha_j^k = 1$  denotes that a  $X^{1/2}$  gate was applied at qubit  $j$  in (clock cycle)  $k$ .

# Computational complexity

- For **universal circuits**,  $p_U(x) = \lambda|Z|^2$  is proportional to the partition function  $Z = \sum_s e^{i\theta H_x(s)}$  of an Ising model  $H_x(s) = h_x \cdot s + s \cdot \hat{J} \cdot s$  with complex temperature  $i\theta (= i\pi/8)$  and **no structure**.
- $Z$  has a strong sign problem:  $Z = \sum_j M_j e^{i\theta E_j}$ ,  $|M_j|$  exponentially larger than  $|Z|$ .
- Worst-case complexity:  $Z$  can not be probabilistically approximated asymptotically with an NP-oracle (is #P-hard). (Fujii and Morimae 2013, Goldbert and Guo 2014).
- Computational complexity conjecture: average case = worst case complexity. There is no structure. (Bremner et. al. 2015).
- Theorem: if  $p_U(x)$  can be classically sampled, then  $Z$  can be approximated with an NP-oracle (Bremner et. al. 2015). **Contradiction**.
- Classical factoring has **no** computational complexity implications.

# Simulation time

<b>% of comm</b>	<b># of sockets</b>	<b># of fused</b>	<b>Avg. time per gate (sec)</b>	<b>Time per Depth-25 (sec)</b>
<b>5 × 4 circuit: 20 qubits, 10.3 gates per level, 17 MB of memory</b>				
0.0%	1	0.00	0.00015	<b>0.039</b>
<b>6 × 4 circuit: 24 qubits, 12.5 gates per level, 268 MB of memory</b>				
0.0%	1	7.01	0.0041	<b>1.294</b>
<b>6 × 5 circuit: 30 qubits, 16.2 gates per level, 17 GB of memory</b>				
0.0%	1	5.64	0.349	<b>141.3</b>
<b>6 × 6 circuit: 36 qubits, 19.5 gates per level, 1 TB of memory</b>				
6.2%	64	5.40	0.76	<b>369.0</b>
<b>7 × 6 circuit: 42 qubits, 23.0 gates per level, 70 TB of memory</b>				
11.2%	4,096	5.54	1.72	<b>989.0</b>

On Edison, a Cray XC30 with 5,576 nodes. Each node is dual-socket Intel® Xeon E5 2695-V2 with 12 cores per socket, 2.4GHz. 64GB per node (32GB per socket). Nodes connected via Cray Aries with Dragonfly topology. (Mikhail Smelyanskiy).

# Conclusions

- We expect to be able to approximately sample the output distribution of shallow random circuits of  $7 \times 7$  qubits with significant fidelity in the near term.
- It is impossible to approximately sample the output distribution of shallow random quantum circuits of  $\approx 48$  qubits with state-of-the-art supercomputers ( $d \sim 25$ ).
- **Quantum supremacy.**
- New method to benchmark complex quantum circuits efficiently.
- Relation to quantum chaos.
- Relation to computational complexity.
- The theory applies to other chaotic systems: chaotic Hamiltonians, commuting quantum circuits, BosonSampling.
- The cross entropy method applies to other sampling problems.