

# Verification of NISQ Devices

...

From Benchmarking to Protocol Verification



# Introduction

# NISQ

- Few qubits (100-200) - even less
- Limited architecture
- Lots of Noise (I mean really... wow)
  - Verification compensates for lack of error correction
- Verification of sampling
- No fault tolerance and in some cases no error correction

# Verification - What do they want?

## Physicists

- Certify the outcome of their simulation (ground state/noise)
- Accurately determine physical properties (entanglement/phase estimation/purity)
- Trust in device as “good” quantum simulator in many situations (benchmarks)

## Industry

- Trust in quantum computer/simulator when involving sensitive/public data
- Assurance that quantum computer/simulator is doing what it should be - efficiency/speed-up?

## Computer scientists

- Verify output of quantum computer is correct (classically intractable)
- Security measures for all situations (best to worst case scenario)
- A bound on trust in your NISQ or UQ device

## The public

- “So, if I use a quantum computer to google something it will give me the results even faster and they’ll be better??”
- Are my transactions secure?
- Can we have better drugs and are they safe?



# Randomized Benchmarking

# What Do You Need And What Can You Get

## Requirements

Any amount of qubits (theoretically)

Set of unitaries/gates that form an exact or approximate unitary t-design from which to sample from.

To efficiently run a number of sequence lengths

Inversion of gates or known basis to measure for final state

## Returns

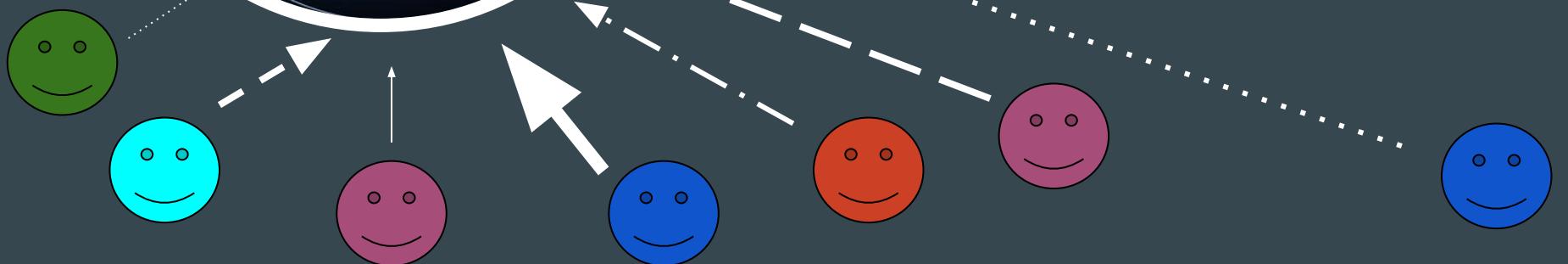
A measure of the average performance of a quantum hardware when running a long quantum information process (partial noise characterisation)

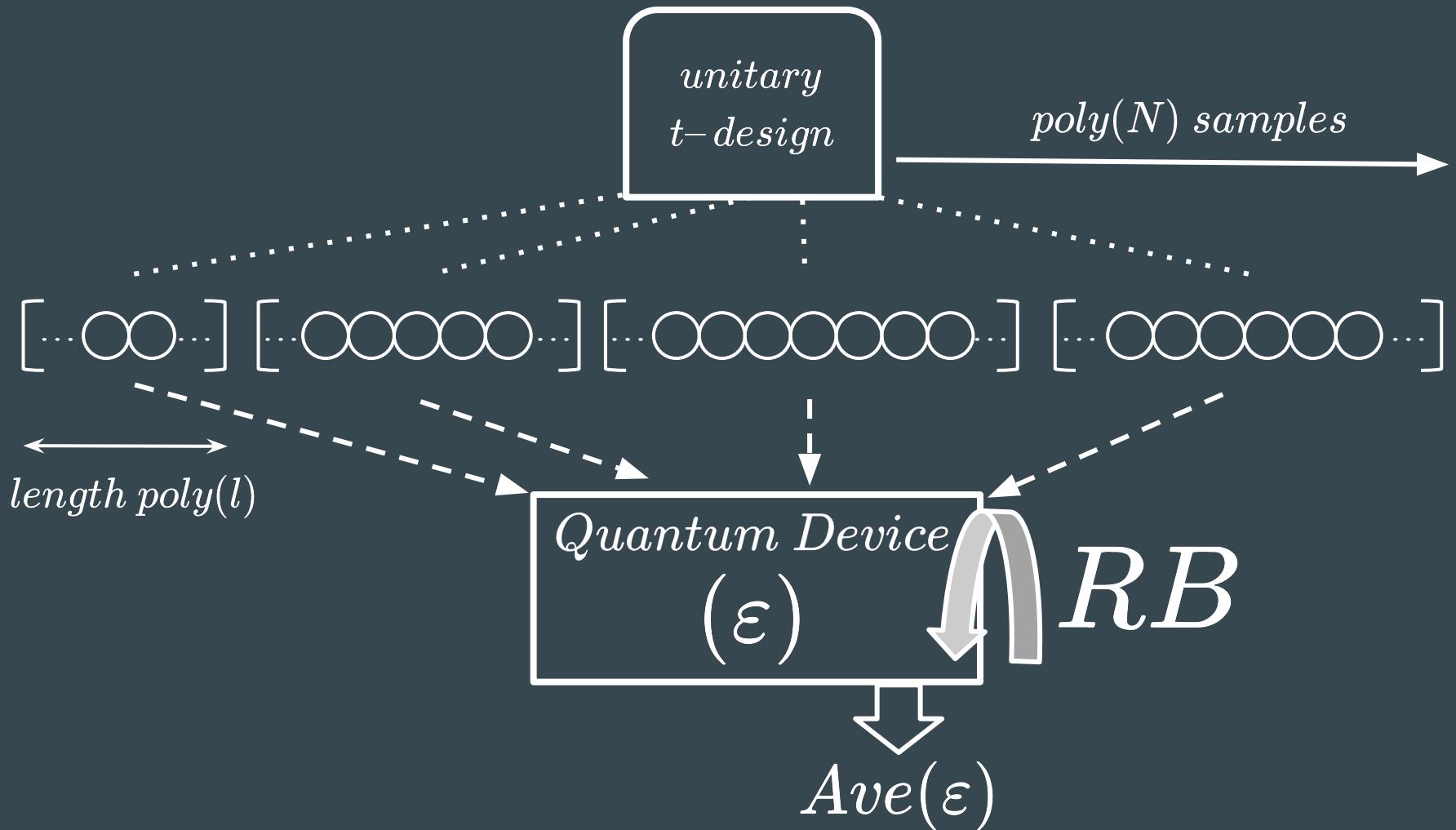
Average error rate of a gateset on your hardware

A measure of a gates performance as a part of a process rather than individually

Incorporates errors from state preparation and measurement

Can I cope?





# Fundamentals

## Twirling

$$\begin{aligned}\bar{\Lambda}(\rho) &= \int_{U(D)} d\mu(U) U \circ \Lambda \circ U^\dagger(\rho) \\ &= \int_{U(D)} d\mu(U) U \Lambda(U^\dagger \rho U) U^\dagger\end{aligned}$$

Average  $\Lambda$  under the composition  
 $U \circ \Lambda \circ U^\dagger$  for unitary operations  
 $U(\rho) = U\rho U^\dagger$  chosen according  
to probability distribution  $d\mu$



If  $d\mu$  is the Haar distribution  
then the twirled channel on  $\rho$   
is a depolarising channel

## Depolarising Channel

$$\bar{\Lambda}(\rho) = p\rho + (1 - p)\frac{1}{D}$$

Strength of channel

# Fundamentals

## Twirling

$$\begin{aligned}\bar{\Lambda}(\rho) &= \int_{U(D)} d\mu(U) U \circ \Lambda \circ U^\dagger(\rho) \\ &= \int_{U(D)} d\mu(U) U \Lambda(U^\dagger \rho U) U^\dagger\end{aligned}$$

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## Depolarising Channel

$$\bar{\Lambda}(\rho) = p\rho + (1 - p)\frac{1}{D}$$

Strength of channel

If  $d\mu$  is the Haar distribution then the twirled channel on  $\rho$  is a depolarising channel

Unitary t-design



# Fidelity

$$F(\Lambda_U, U) = F(U|\psi\rangle\langle\psi|U^\dagger, \Lambda_U(|\psi\rangle\langle\psi|))$$

$$F(\Lambda_U, U) = F(\Lambda_{U,e}, I) \longrightarrow F(\int_{Haar} \Lambda_{U,e}, I)$$

$$\Lambda_U(X) = \sum_i A_i X A_i^\dagger$$

$$\Lambda_U(X) = \sum_i (A_i U^\dagger) U X U^\dagger (U A_i^\dagger)$$

$$\Lambda_U = \Lambda_{U,e} \circ U$$

$$\Lambda_{U,e} = \sum_i A_i U^\dagger \otimes U A_i^\dagger$$

# Fidelity

$$F(\Lambda_U, U) = F(U|\psi\rangle\langle\psi|U^\dagger, \Lambda_U(|\psi\rangle\langle\psi|))$$

$$F(\Lambda_U, U) = F(\Lambda_{U,e}, I) \longrightarrow F(\int_{Haar} \Lambda_{U,e}, I)$$

$$\int_U d\mu(U) F(\Lambda_U, U) = \int_U d\mu(U) F(\Lambda_{U,e}, I)$$



$$(1 - \frac{D-1}{D}) + \frac{D-1}{D}(1 - \overline{p_d})$$

$$\Lambda_U = \Lambda_{U,e} \circ U \longrightarrow \Lambda_{U,e} = \Lambda_U \circ U^{-1}$$

# Method

$$\rho = |\psi\rangle\langle\psi|$$

REPEAT

$$\Lambda_{U_{Tot}}^{-1} \Lambda_{U_m} \dots \Lambda_{U_1} \longrightarrow M = \{E, 1 - E\}$$

Average survival probability

$$Tr(ES_m(\rho))$$

- Average each survival probability over number of sequences sampled at that length: average survival probability over all possible sequences at that length
- Do this for varying sequences - where all unitaries are sampled from a unitary t-design.

$$P_m = A + (B + Cm)p^m$$

$$r = \frac{D-1}{D}(1-p)$$

# Verified/Secure?

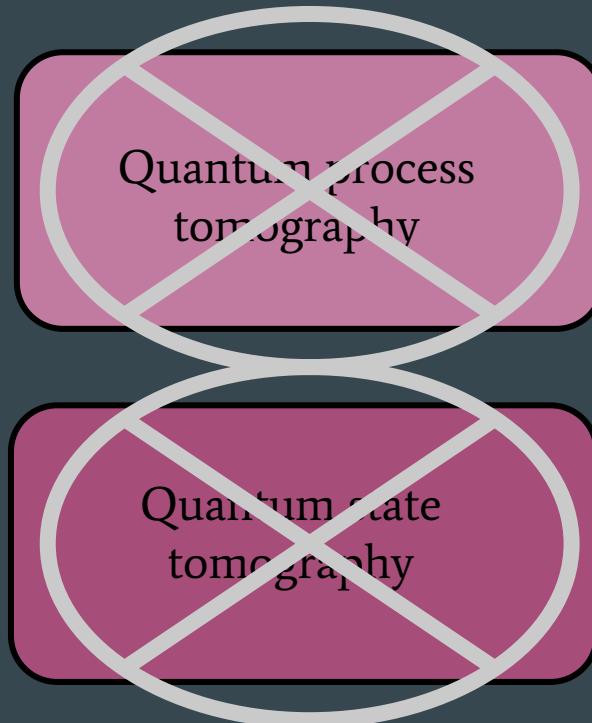
Random processes not specific algorithms - correct outcome of computation **not verified** with this technique

The “server” and “verifier” know the initial state of the system, the random processes run on the device and the measured output - **not secure**

If your specific algorithm were hidden in the random processes somehow, could we get a measure of the average error rate for that process on the hardware without the “server” knowing what the algorithm was?

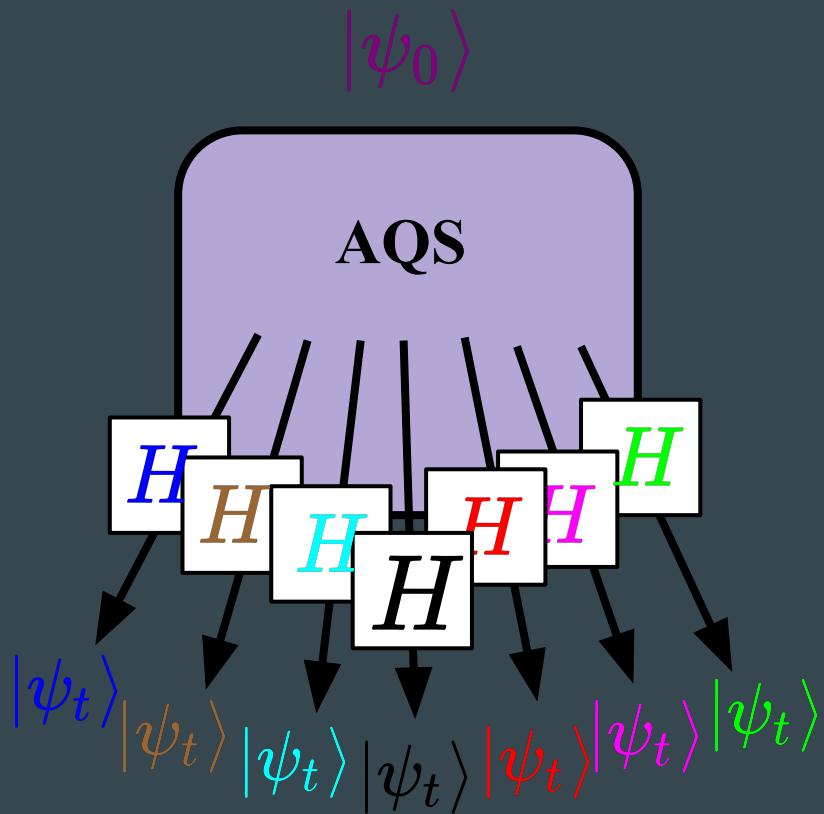
# In the analog setting - why is this interesting ?

Not efficiently scalable



Motivation To develop a method for testing and  
incorporate state preparation and  
measurement errors

# Programmable analog quantum simulators



Tunable

Reproducible

For:

Problems that require being able to run a whole class of hamiltonians in a reproducible way.

Way to test/certify such a simulator?

# In the analog setting - RB method

$$\left\{ H_s + \begin{matrix} k \\ \text{Disorder} \end{matrix} \right\} = \left\{ H_k \right\}$$

$$\left\{ U_k = e^{-iH_k t} \right\} \xrightarrow{\text{Imperfectly implemented: } \Lambda_{U_k} = \varepsilon \circ U_k} \text{Set to sample unitaries}$$

# In the analog setting - RB method

Same as standard RB but :

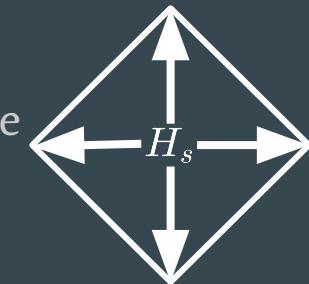
- Unitaries are time-evolution operators sampled from set generated from native gates of system
- Each unitary is systematically inverted (for now) rather than one single deterministic inversion operator

# Generating a unitary t-design from Hamiltonian

Non trivial problem

Generate disorder around static Hamiltonian - break symmetry enough to generate a unitary t-design - (disorder potential + interaction term)

- Product of those generated will eventually span unitary space
- For 2-design can compare second moment of Haar measure with second moment of unitaries generated : basically compare eigenvalues - should be two max with 1 and 0's everywhere else



# Verification with randomized benchmarking?

- Can we get an average error rate for a specific quantum algorithm?
- Need it to appear random, or be hidden within a random sequence

$$\left\{ \begin{matrix} U_{alg} & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & U_t \\ U_t & \cdot & \cdot & \cdot \end{matrix} \right\}$$

Embed specific algorithm sequence within sequence of random unitaries from unitary t-design.

- Build unitary t-design around specific algorithm?



# Quantum Benchmark (company)

# Claims

“The True-Q(™) Validation software system *accurately* validates the Quantum Capacity of *any* quantum hardware platform to execute any quantum circuit for *any* user-supplied problem or application.”

“Validates the capacity of *any* quantum hardware platform to perform *any* user-supplied algorithm to *any* user-specified precision”

“True-Q(™) Design is a scalable solution for optimizing hardware design and quantum computing performance.”

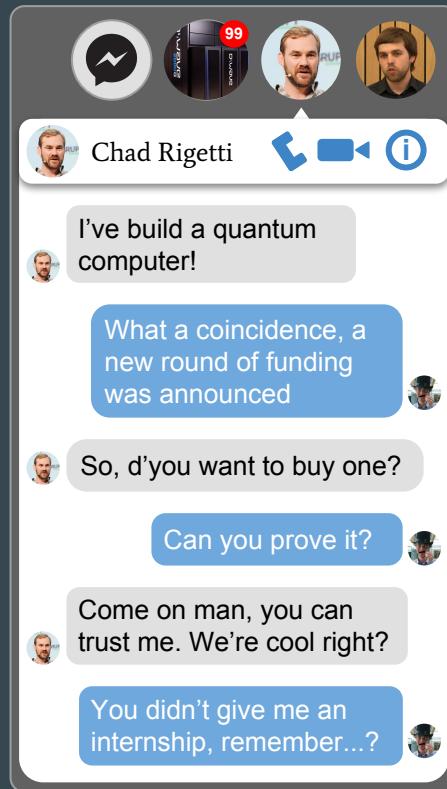
# How do they achieve this?

- ***Randomized Benchmarking***: accurate and precise error characterization of elementary quantum gates
- ***Cycle Benchmarking***: scalable error characterization of arbitrary parallelized gate cycle and universal (polynomial-depth) quantum circuits
- ***Scalable Error Reconstruction***: detailed error reconstruction across the quantum processor to find error correlations and optimize hardware design and performance of quantum error correcting codes
- ***Randomized Compiling***: efficient run-time error suppression for arbitrary applications
- ***Quantum Capacity***: high-precision performance validation for arbitrary applications



# Hypothesis Testing

# The Setting



# Superiority Null Hypothesis

*The set of samples which I have in my possession were drawn from a distribution produced by a classical computer in polynomial time*

Unlike traditional experiments this amounts to the *nonexistence* of something. Hence we need some theoretical tools to guide us

# Boson Sampling

Constant-Depth Quantum Circuits

*extended clifford circuits*

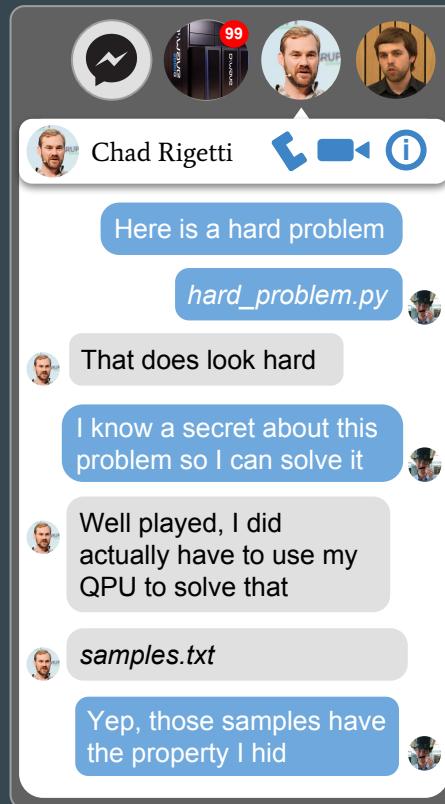
*IQP*

*one-clean-qubit*

# QAOA

Ball Permutations

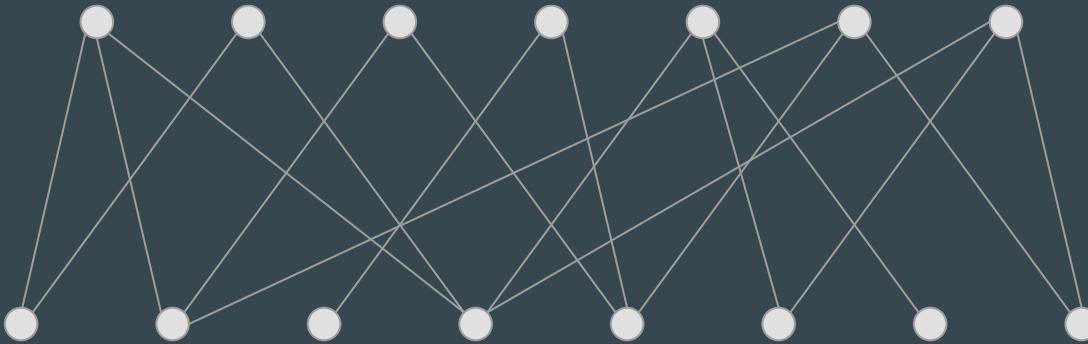
# One Possible Option



# Some Components of the Hypothesis Test to Extract

1. A reason Chad must use a quantum computer
  - Hard computational problem
2. Property of the outcome, which is “highly correlated” to the outcome, to check
  - The small hidden problem should be solvable and indicative of the larger problem
3. A backdoor that helps us check property
  - A smaller problem should be hard to uncover
4. Means to implement on NISQ devices
  - Let’s figure something out for IQP... Why not?

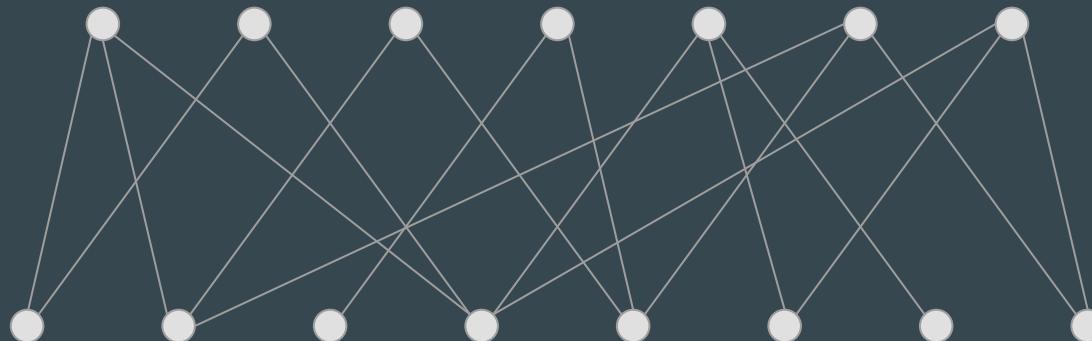
# An Example



# An Example

“Output qubits”

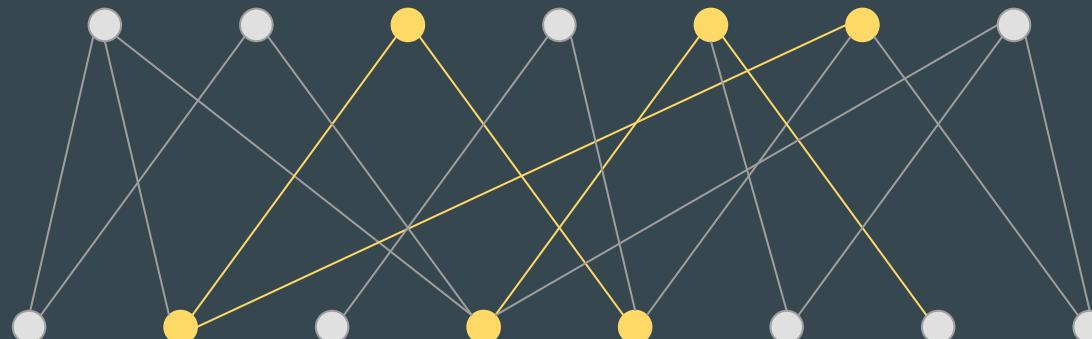
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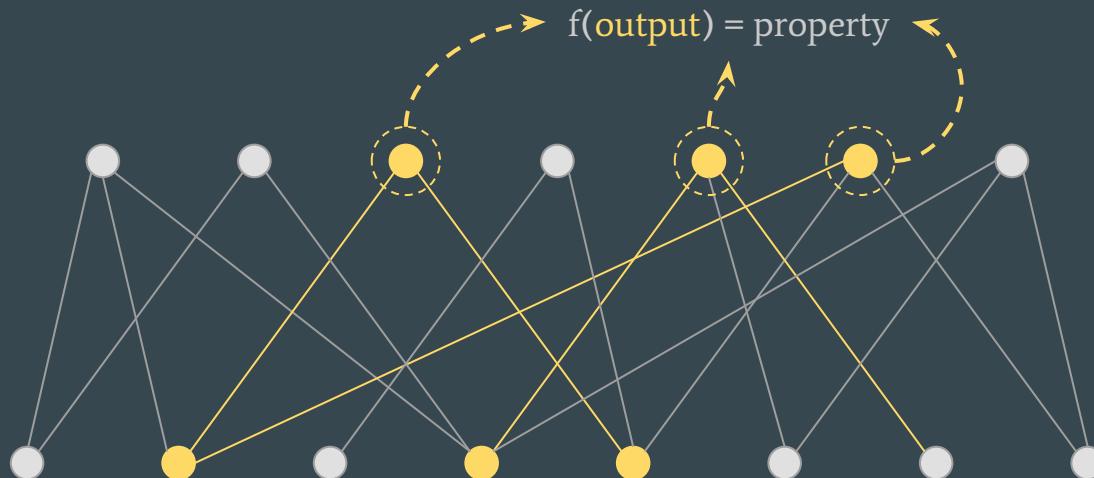
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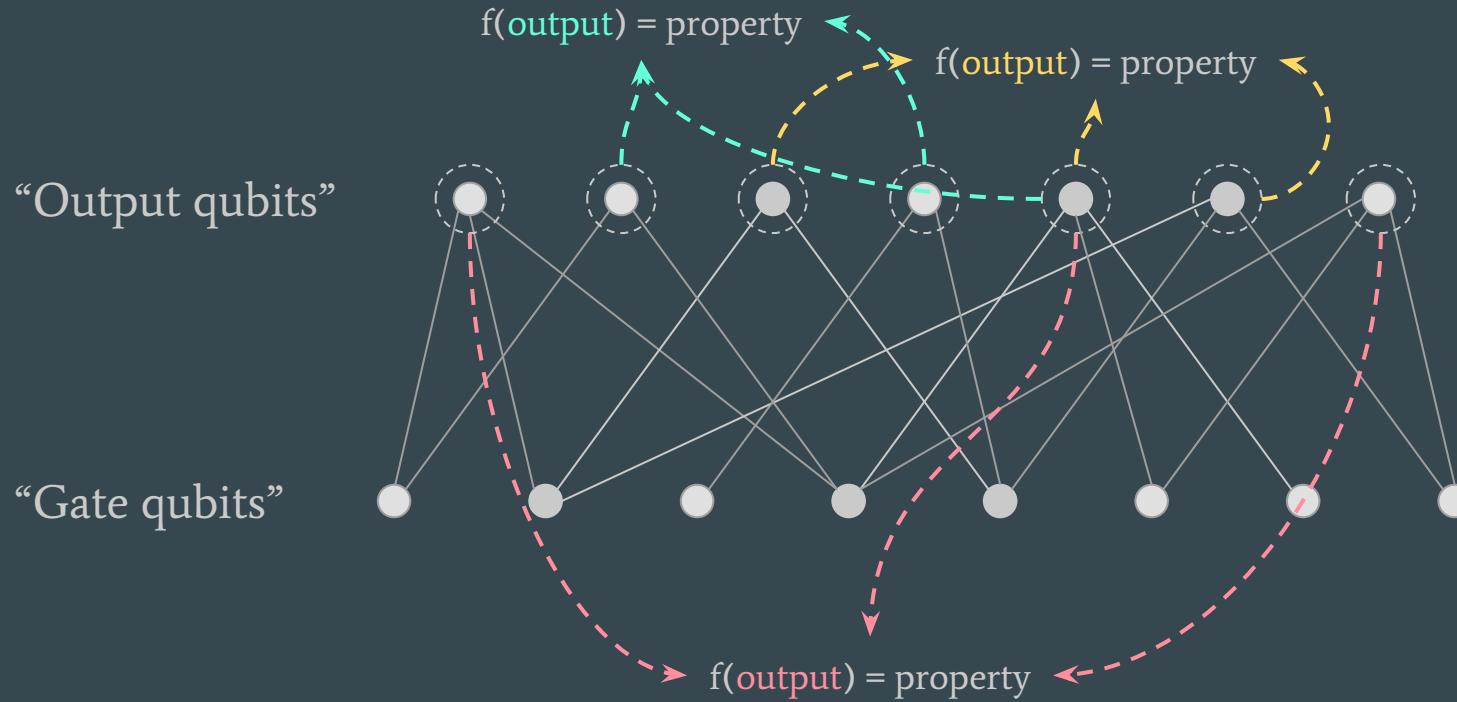
“Output qubits”

“Gate qubits”



$f(\text{output}) = \text{property}$

# Chad's View



# Chad's View

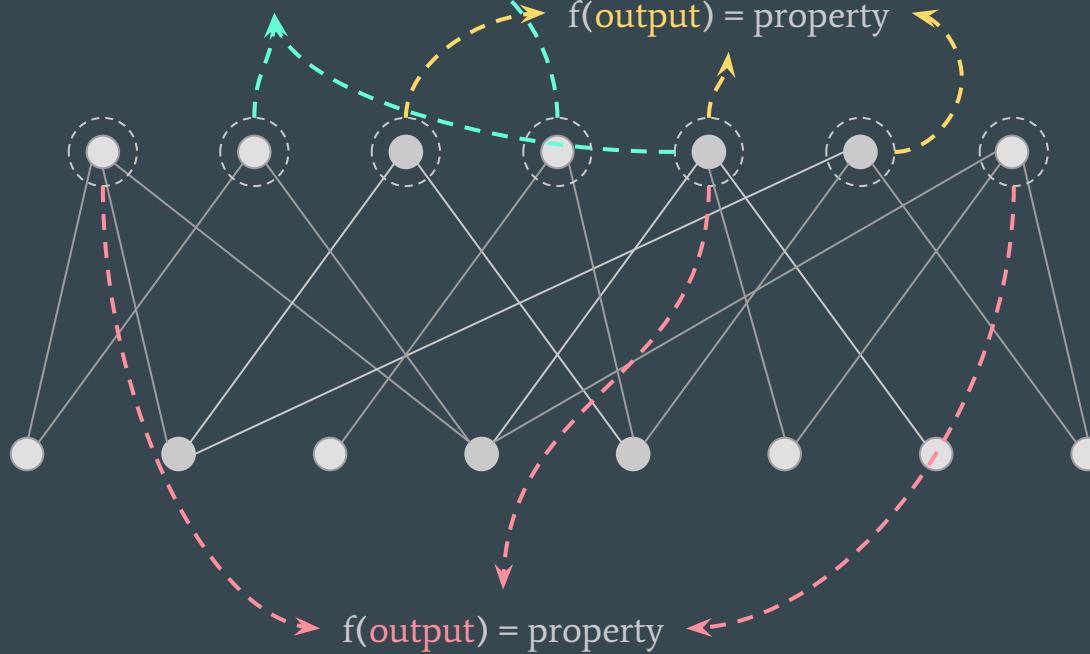
“Output qubits”

“Gate qubits”

$f(\text{output}) = \text{property}$

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?!?!?!?!



# It Meets The Requirements?

1. A reason Chad must use a quantum computer
  - o It looks like a big IQP computation to him
  - o Cannot reproduce classically as hiding is good
2. Property of the outcome, which is “highly correlated” to the outcome, to check
  - o The property of the hidden graph is fixed so can be checked
  - o Its embedding in the larger graph makes it highly correlated
3. A backdoor that helps us check property
  - o You know where the small problem is!
4. Means to implement on NISQ devices
  - o IQP is easier to implement than BQP

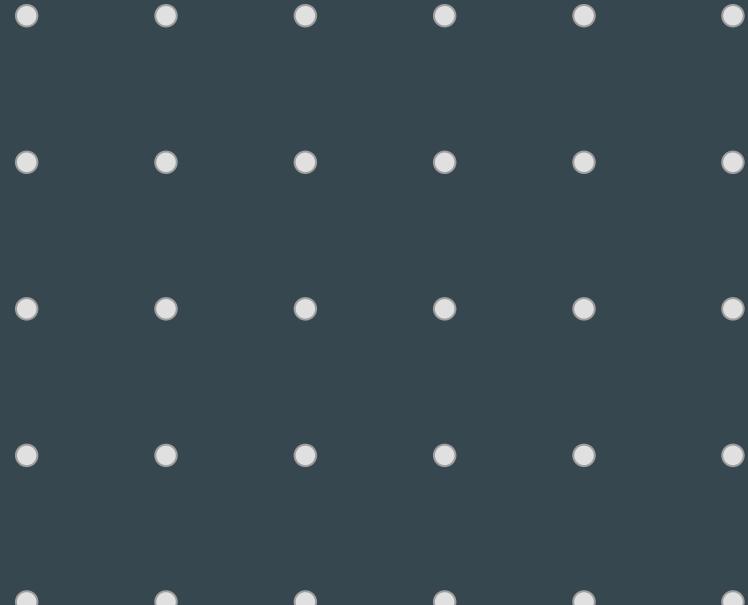
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**FORBIDDEN**
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# Random Circuit

For example:

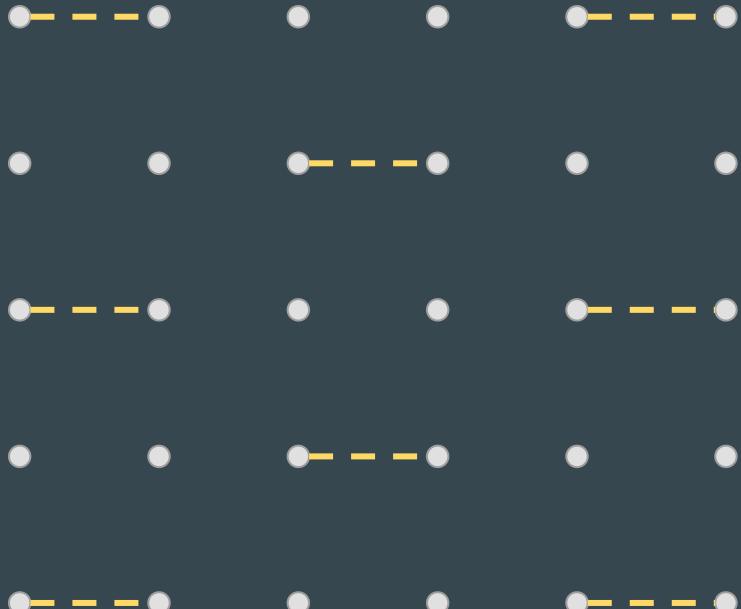
- 1 Cycle of Hadamard gates
- 2 For d clock cycles:
  - 3 Apply CZs
  - 4 If no CZ applied
    - 5 If no random gate acted yet
    - 6 Apply T
  - 7 Else
  - 8 Apply gate different from previous



# Random Circuit

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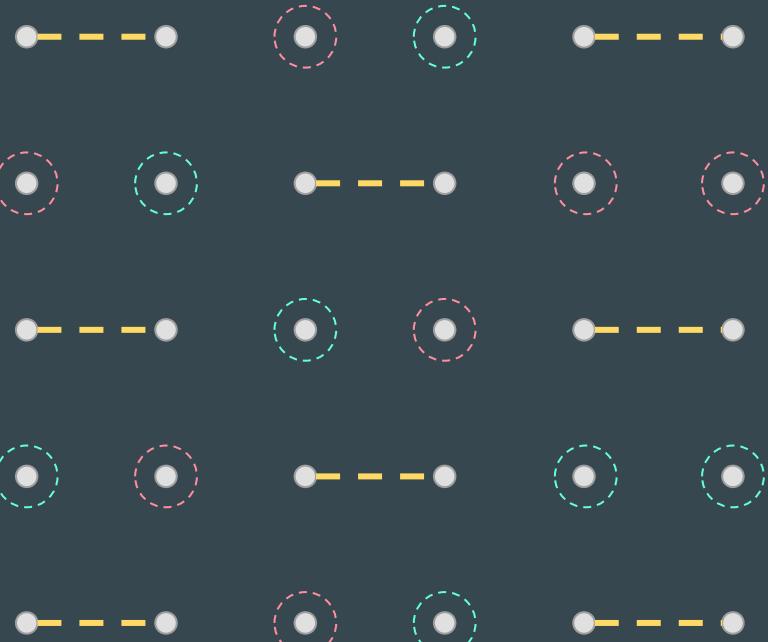
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# Heavy Output Generation

*Given as input a random quantum circuit  $C$ , generate output strings  $x_1, \dots, x_k$  at least  $\frac{2}{3}$  fraction of which have greater than median probability in  $C$ 's output distribution.*

Relational problem which can be verified in classical exponential time by calculating ideal probabilities

# Under what assumption is HOG classical hard

Quantum Threshold assumption:

*There is no polynomial time classical algorithm that takes a description of a random quantum circuit  $C$ , and that guesses whether  $|\langle 0^n | C | 0^n \rangle|^2$  is greater or less than the median of the values of  $|\langle 0^n | C | x \rangle|^2$ , with success probability at least  $\frac{1}{2} + \Omega(\frac{1}{2^n})$  over the choice of  $C$ .*

# Quantum Threshold Assumption

- There is simple reduction
  - HOG is not hard  $\Rightarrow$  there exists polynomial-time algorithm to find high probability outputs  $\Rightarrow$  one can use this algorithm to guess  $|\langle 0^n | C | 0^n \rangle|^2 \Rightarrow$  QUATH does not hold
- Despite similarity between HOG and QUATH, importantly it is not a relational problem and does not refer to sampling.
- Justified through rather flimsy reasoning

# How Does This Relate to Our Comments From Before

1. A reason Chad must use a quantum computer
  - o If QUATH hold then he'll have to
2. Property of the outcome, which is “highly correlated” to the outcome, to check
  - o Did he meet the conditions of the HOG problem?
3. What price did I pay for removing the backdoor that helps us check property
  - o Actually it takes exponential time to check this... You just have to brute force it
4. Means to implement on NISQ devices
  - o Random circuits are \*THE\* NISQ device ... google it

# Cross Entropy Difference

Measure quality as the difference from uniform classical sampler

$$\Delta H(p_A) = \sum_j \left( \frac{1}{N} - p_A(x_j|U) \right) \log \frac{1}{p_U(x_j)}$$

- Unity for ideal implementation
  - Output entropy equal to Porter-Thomas distribution
- Zero for uniform distribution

Achiever supremacy in range:

$$1 \geq \Delta_{\text{cross-entropy}} > C$$

# A Classical Computer Cannot Pass a Cross-Entropy Test?

Approximating cross entropy difference (probably) requires explicitly calculating probabilities

1. This means  $C = 0$  for large circuit
2. Also means we cannot measure cross entropy difference for large circuits

\*whispers\* we can probably just extrapolate \*whispers\*

It is argued that approximating the probabilities is hard and a weaker assumption than QUATH

# How Does This Relate to Our Comments From Before

1. A reason Chad must use a quantum computer
  - Producing Porter-Thomas distributions requires a quantum computer
2. Property of the outcome, which is “highly correlated” to the outcome, to check
  - Can cross-entropy benchmark it
3. What price did I pay for removing the backdoor that helps us check property
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**Spoiler! It doesn't work anyway**

# What Have We Learned

- Hypothesis tests are used to prove “quantumness”
- They require a property which should be checked that is “highly correlated” to the hard problem being implemented
- This highly correlated property is sort of the key here

# What Have We Learned

We've learned  
we should wear  
sunscreen



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# Future Work

- Does not seem to be a reason to restrict to Random Circuits
  - Or maybe...
  - Random circuits are very flexible
- Can we use these hypothesis tests as a kind of “*meaningful*” verification
- What do hypothesis test teach us about limits of classical computers
  - Where will we see superiority
- Can the IQP random circuits be restricted to square lattices nicely
  - Can we combine runtime of IQP into Random circuit NISQness



# Building Trust For Quantum States

# Quantum State Tomography

- Reconstructing the density matrix of a quantum state (output of an experiment)
- Many measurements and various measurement settings
- Scales exponentially in the number of subsystems (accounting for all correlations)
- *Independently and Identically Distributed* (IID) assumption

# Quantum State Certification

Target state  $\rho$ , direct fidelity estimation

$$1 - \sqrt{F(\rho, \sigma)} \leq \text{Tr}(|\rho - \sigma|) \leq \sqrt{1 - F(\rho, \sigma)}$$

IID assumption:  $\sigma^N = \sigma^{\otimes N}$

# Quantum State Verification

Target state  $\rho$ ,

$$F(\rho, \sigma) \geq 1 - \epsilon \text{ with probability greater than } 1 - \delta$$

where  $N = \text{poly}(\frac{1}{\epsilon}, \frac{1}{\delta})$

No IID assumption!

# Quantum State Verification Beyond Tomography

Target state  $\rho$ ,

$$F(\rho^{\otimes m}, \sigma^m) \geq 1 - \epsilon \text{ with probability greater than } 1 - \delta$$

where  $N = \text{poly}(m, \frac{1}{\epsilon}, \frac{1}{\delta})$

No IID assumption!

# What About CV?

Infinite Fock basis:  $\{|n\rangle\}_{n \in \mathbb{N}}$

$$\rho = \sum_{k,l=0}^{\infty} \rho_{kl} |k\rangle \langle l|$$

- We are not going to verify all of it: energy cutoff

$$\rho \approx \sum_{k,l=0}^E \rho_{kl} |k\rangle \langle l|$$

# CV Quantum State Tomography

- Finite support over the Fock basis assumption
- IID assumption

Estimating  $\sigma_{kl}$  for  $k, l \leq E$



$$\sigma_{\leq E}^{\otimes N}$$

# CV Quantum State Certification

- Energy test
- IID assumption

Estimating  $F(\rho, \sigma)$  or  $\text{Tr}(A\sigma)$  efficiently



$$\sigma^{\otimes N}$$

# CV Quantum State Verification

- Refined energy test
- No assumptions

Estimating  $F(\rho^{\otimes m}, \sigma^m)$  efficiently

(Proof using De Finetti theorem)



$\sigma^N$

# Outlook

- Extending crypto techniques to CV (no obvious twirling lemma)
- More flexible definitions of security: different measures, robust definitions
- Tailored protocols: trading efficiency and security



# Thanks!