

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)

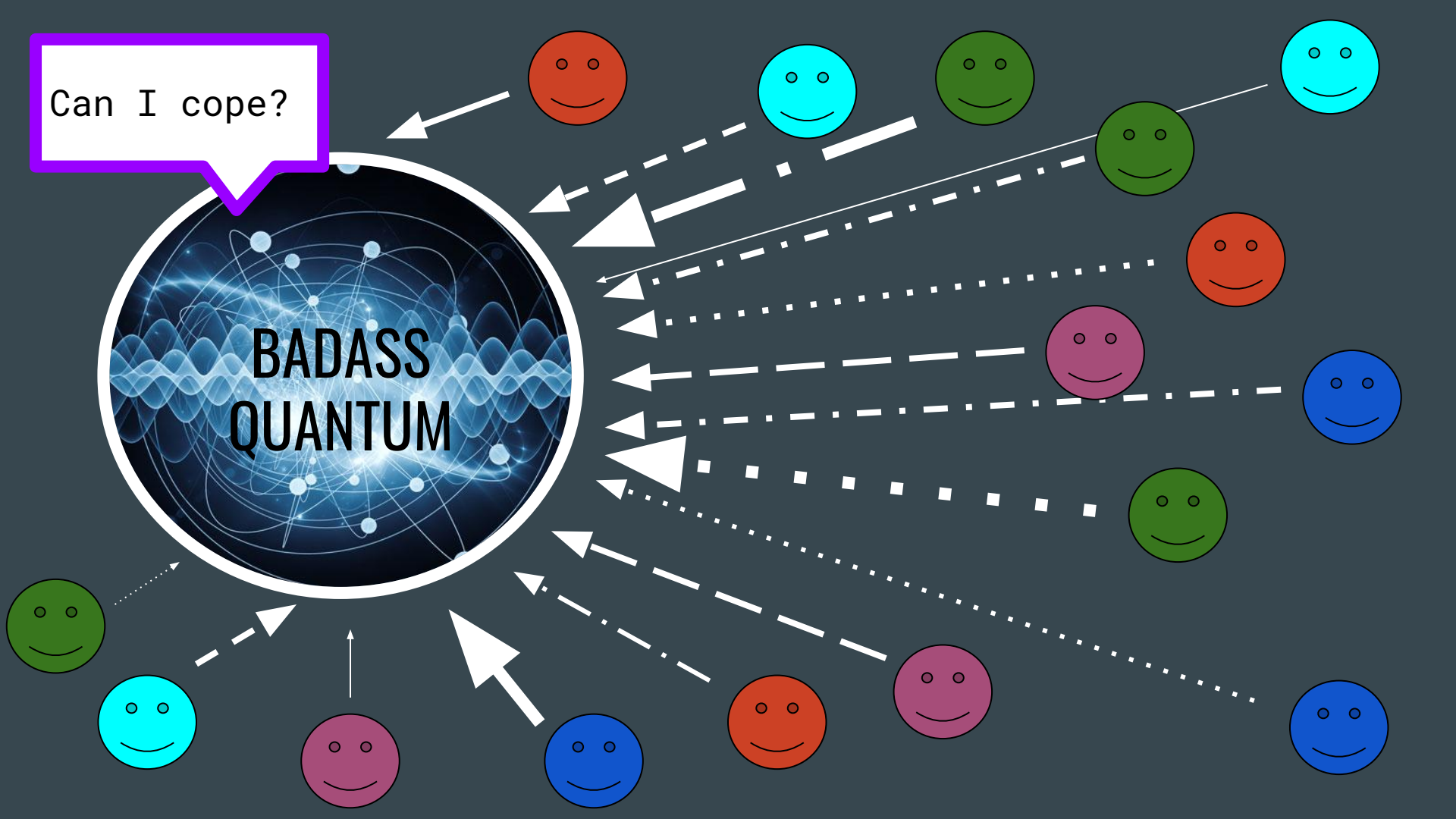
Average error rate of a gateset on your hardware

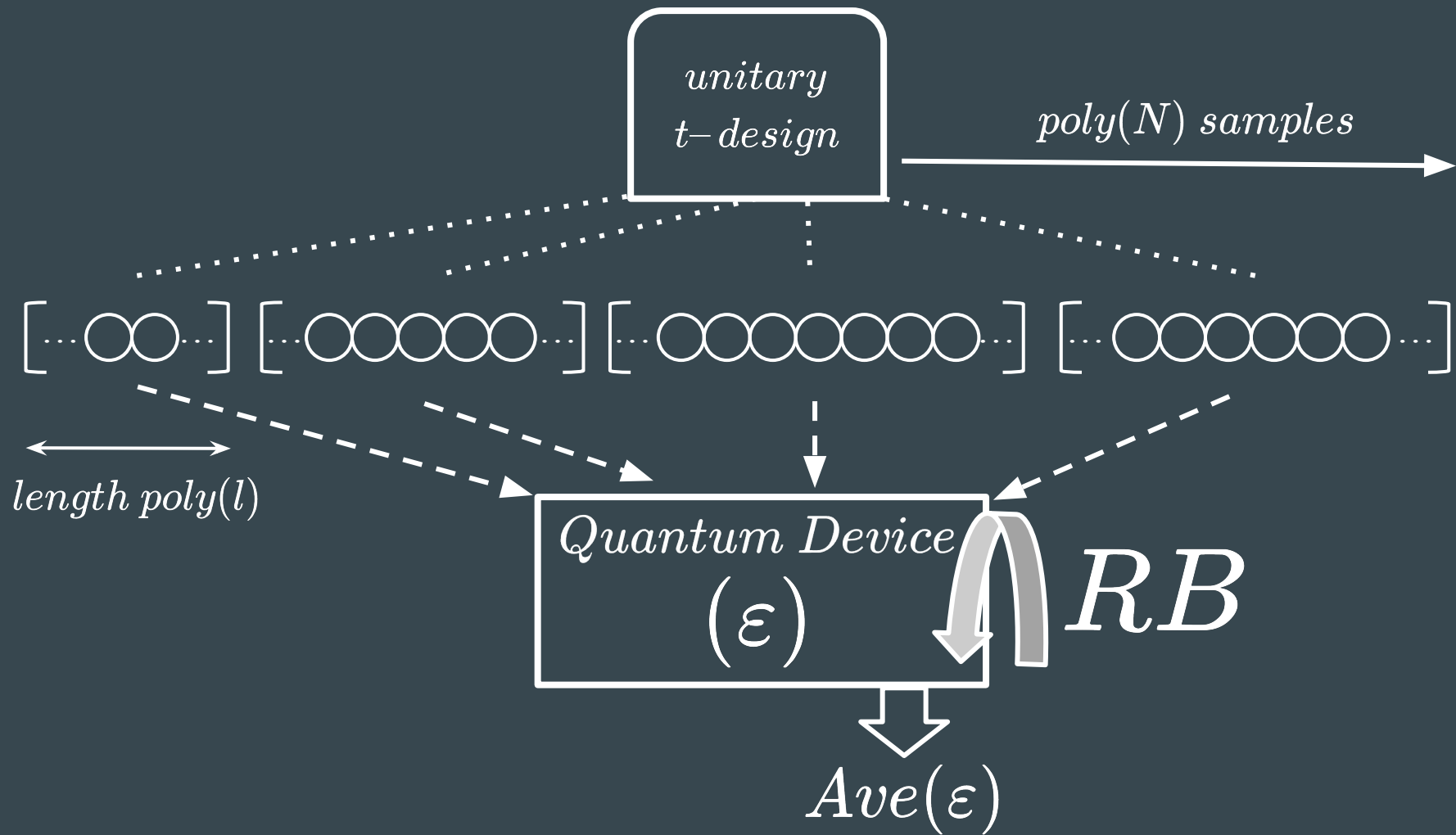
A measure of a gates performance as a part of a process rather than individually

Incorporates errors from state preparation and measurement

Can I cope?

**BADASS
QUANTUM**





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 Λ 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 ρ 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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Unitary t-design

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



Depolarising Channel

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

Strength of channel

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

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

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

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

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

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

Method

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

$$\Lambda_{U_{Tot}}^{-1} \Lambda_{U_m} \dots \Lambda_{U_1} \xrightarrow[\text{Average survival probability}]{\text{REPEAT } M = \{E, 1 - E\}} \text{Tr}(E S_m(\rho))$$

- 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

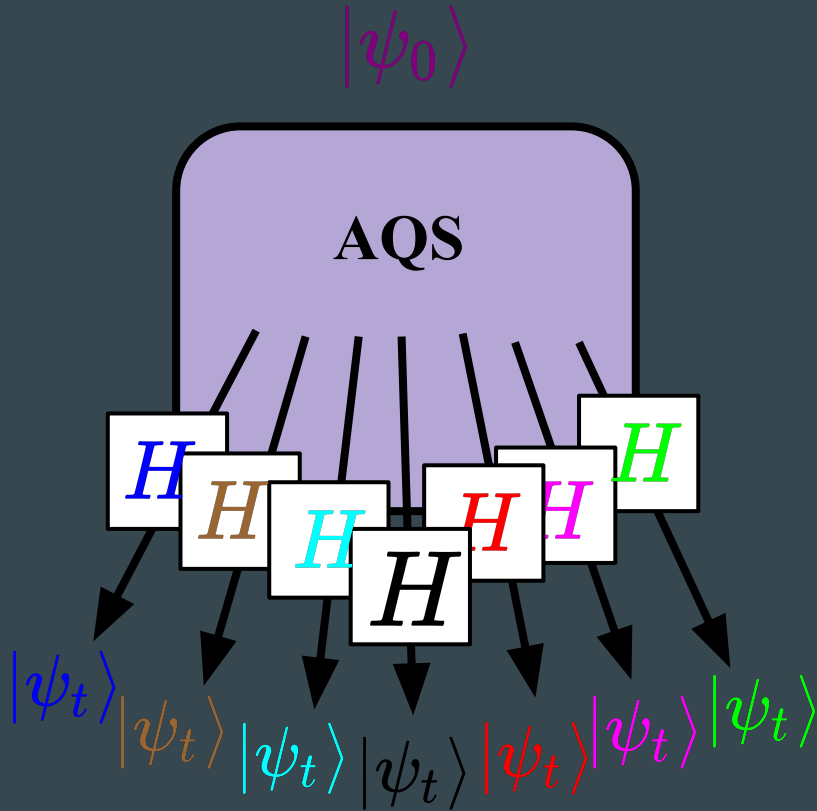
~~Quantum process
tomography~~

~~Quantum state
tomography~~

~~Direct Fidelity
Estimation~~

Motivation: To develop a method for testing and analog quantum simulators that goes beyond the limitations of current techniques

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

$$\left\{ \boxed{H_s} + \begin{array}{c} \triangle \\ k \\ \text{Disorder} \end{array} \right\} = \left\{ H_k \right\}$$

Set to sample unitaries

$$\left\{ U_k = e^{-iH_k t} \right\}$$

Imperfectly implemented: $\Lambda_{U_k} = \varepsilon \circ U_k$

In the analog setting - RB method

Same as standard RB but :

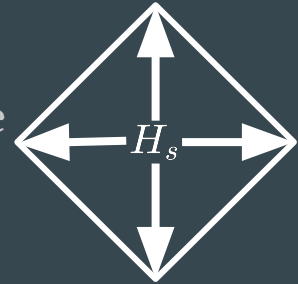
- Unitaries are time-evolution operators sampled from set generated from native gates of system
- Each unitary is systematically inverted (for now) rather than one single deterministic inversion operator

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)

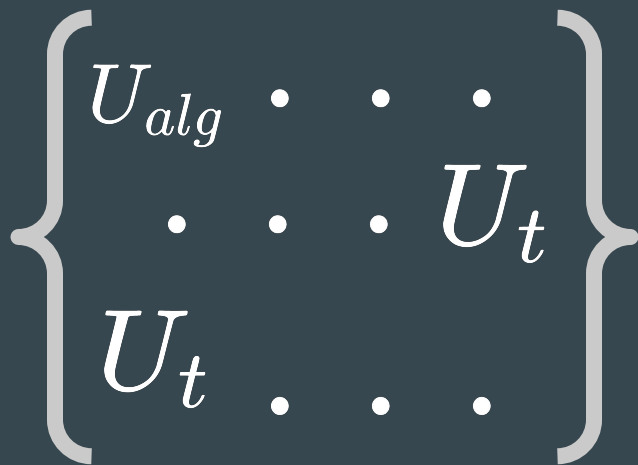
- Product of those generated will eventually span unitary space



- 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



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.”

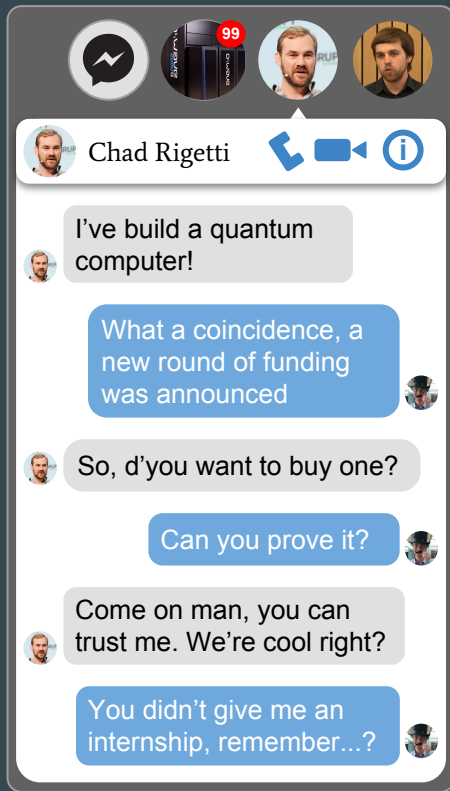
How do they achieve this?

- ***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

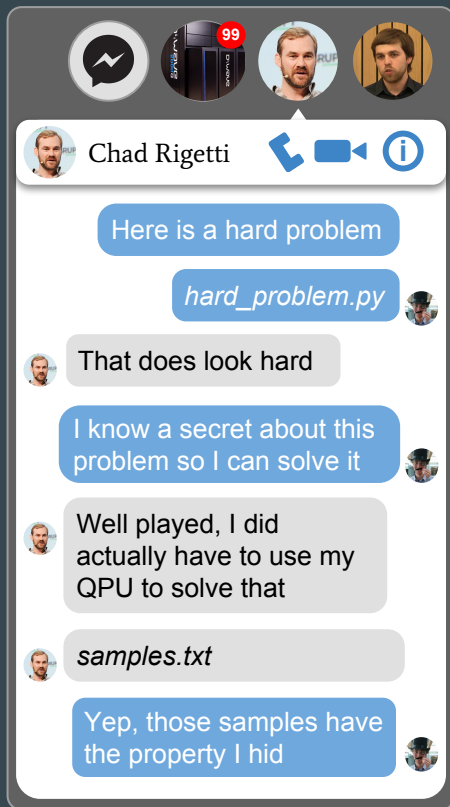
IQP

one-clean-qubit

QAOA

Ball Permutations

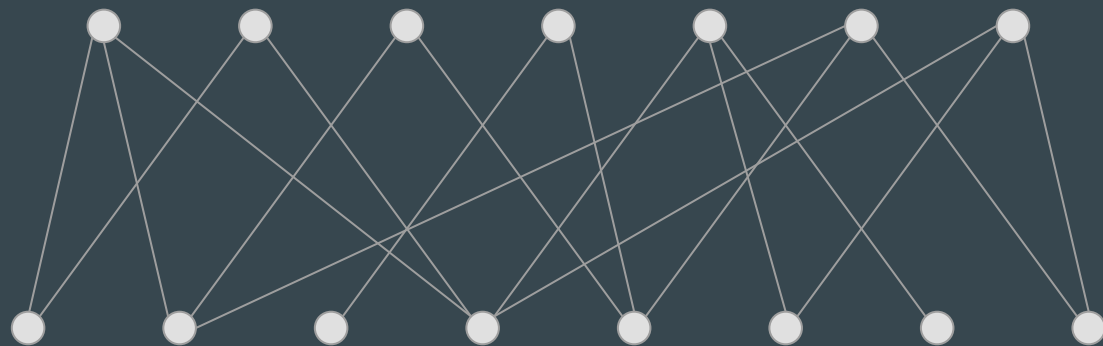
One Possible Option



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?

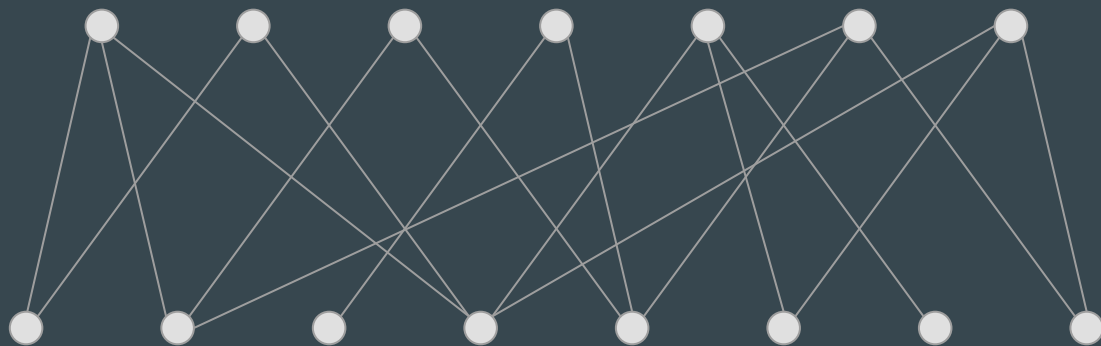
An Example



An Example

“Output qubits”

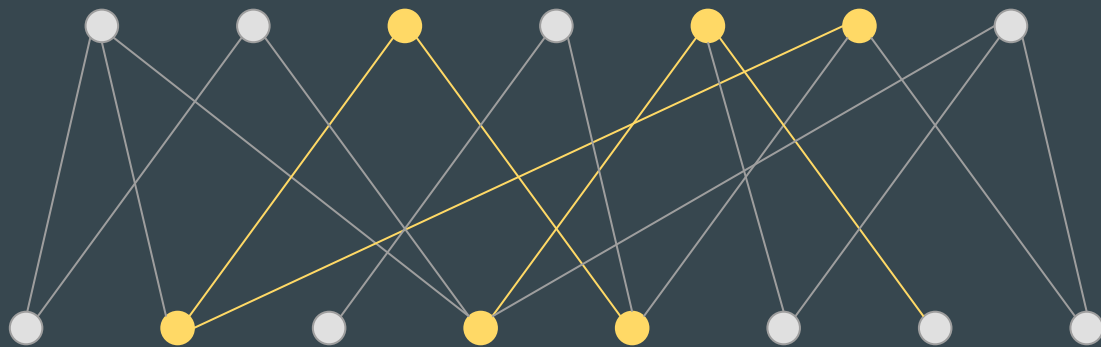
“Gate qubits”



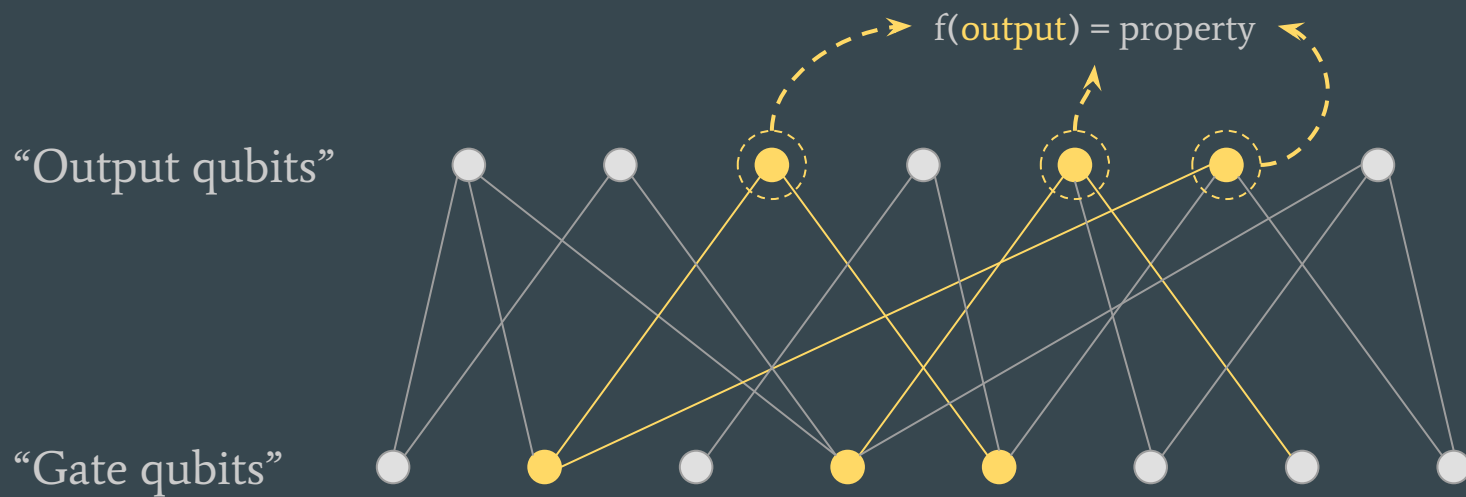
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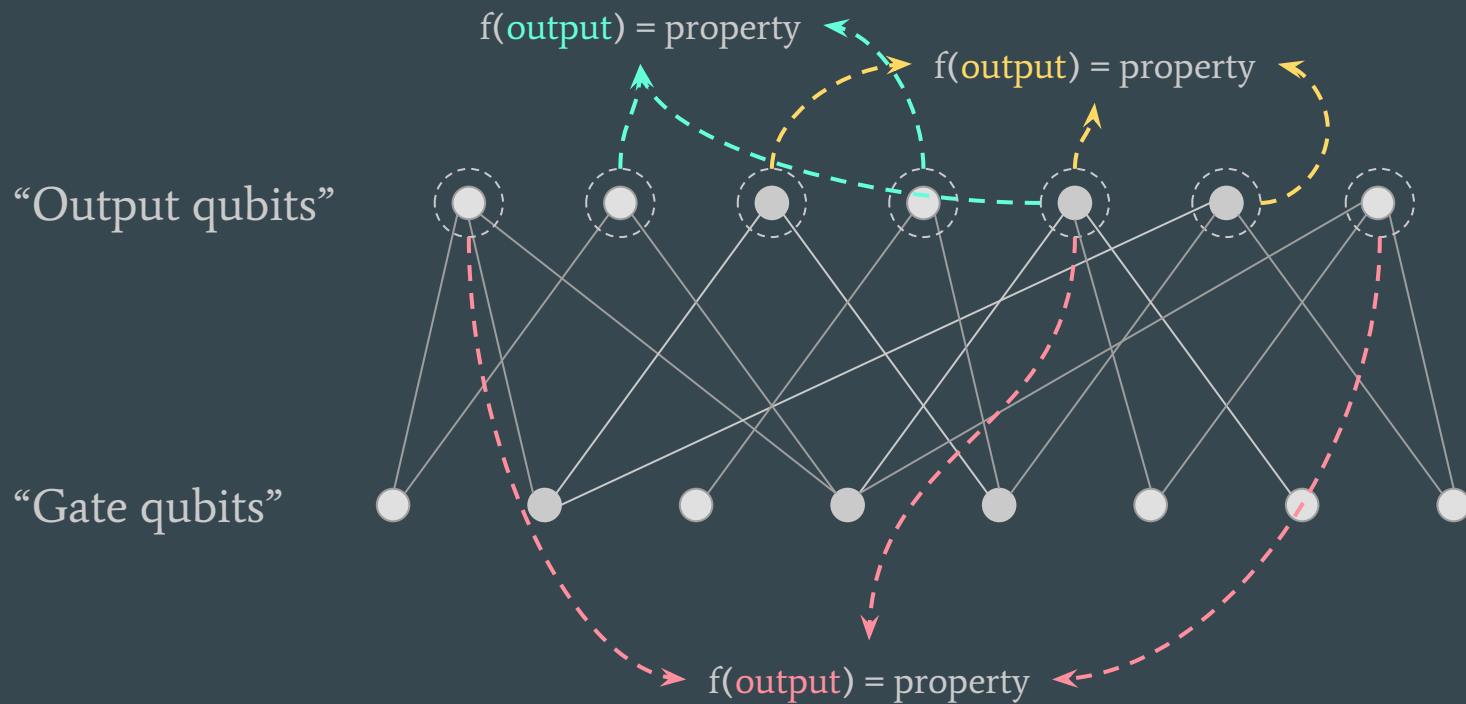
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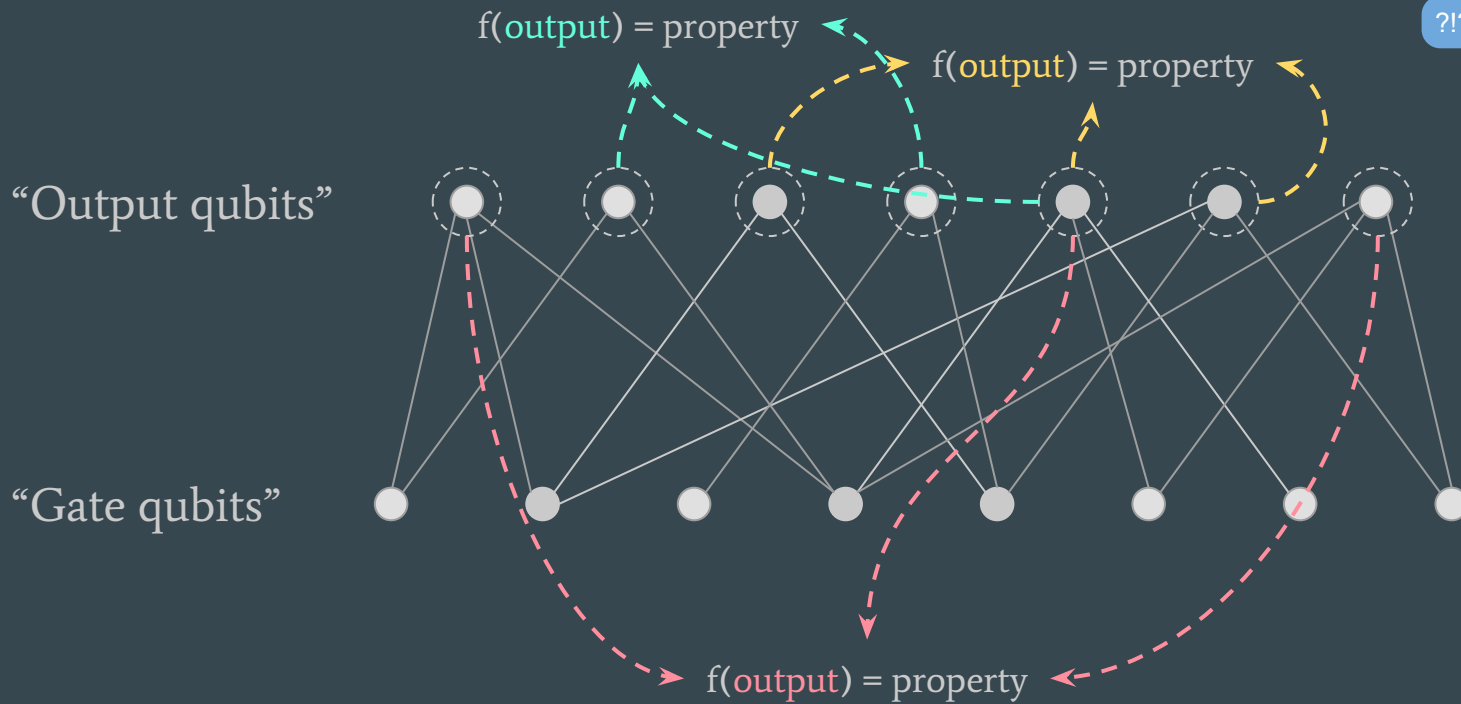
An Example



Chad's View



Chad's View



?!?!?!?!?!?!?!?!?!?!



It Meets The Requirements?

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

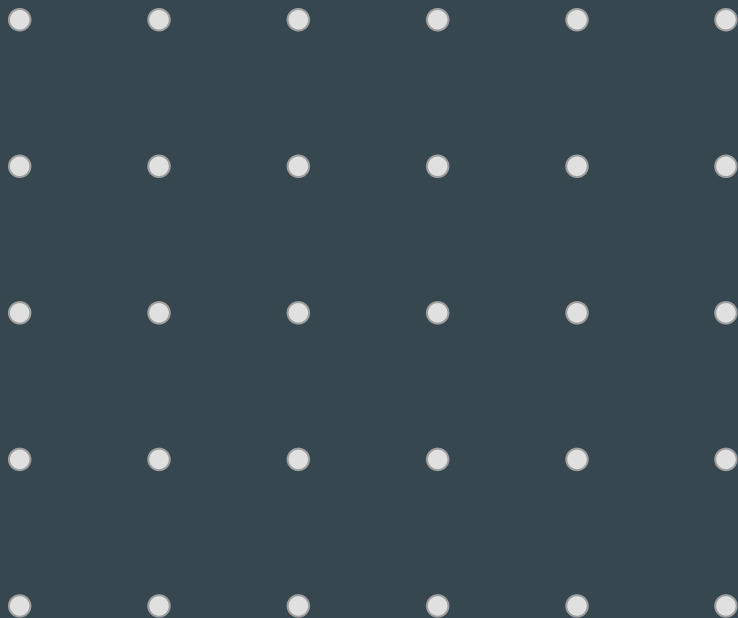
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Random Circuit

For example:

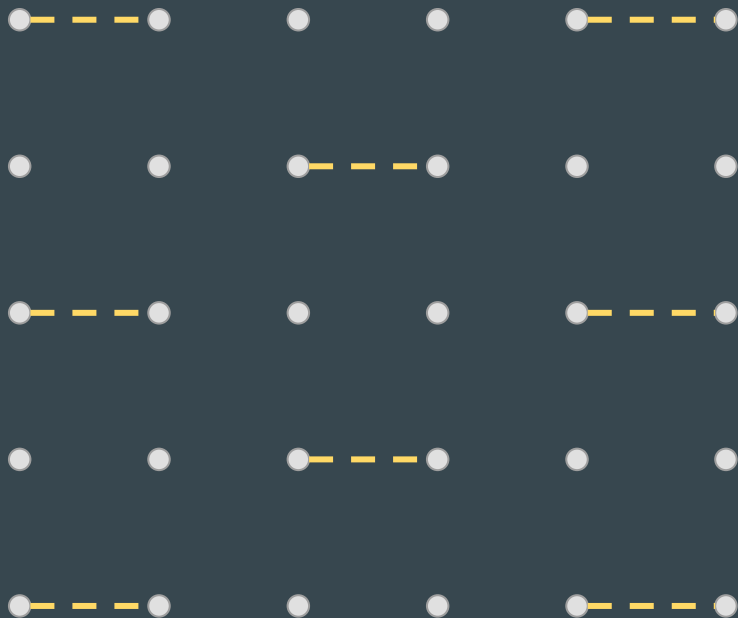
- 1 Cycle of Hadamard gates
- 2 **For** d clock cycles:
- 3 Apply CZs
- 4 **If** no CZ applied
- 5 **If** no random gate acted yet
- 6 Apply T
- 7 **Else**
- 8 Apply gate different from previous



Random Circuit

For example:

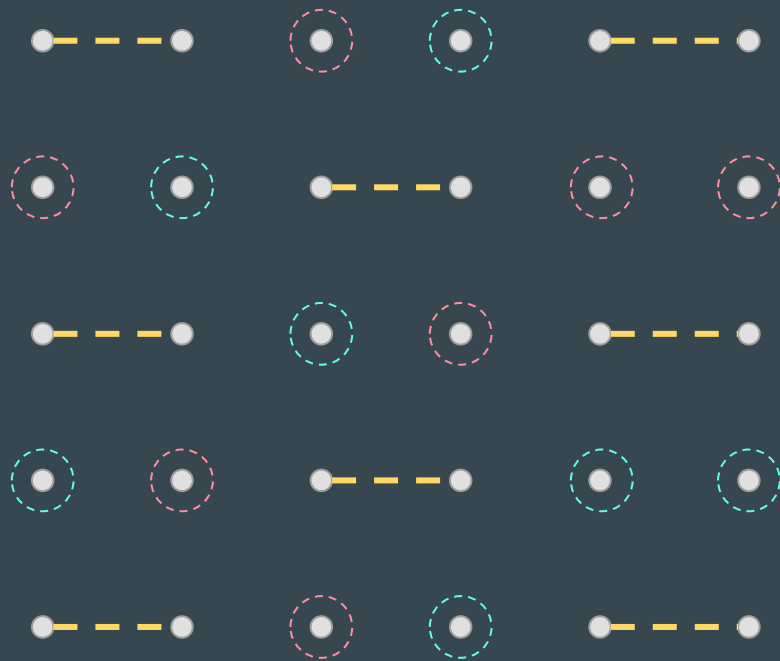
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Random Circuit

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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
 - If QUATH hold then he'll have to
2. Property of the outcome, which is “highly correlated” to the outcome, to check
 - Did he meet the conditions of the HOG problem?
3. What price did I pay for removing the backdoor that helps us check property
 - Actually it takes exponential time to check this... You just have to brute force it
4. Means to implement on NISQ devices
 - 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



- Hypothesis tests are used to prove “quantumness”
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- This highly correlated property is sort of the key here

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 ρ , 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 ρ ,

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

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

No IID assumption!

Quantum State Verification Beyond Tomography

Target state ρ ,

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

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 σ_{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)



σ^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!