

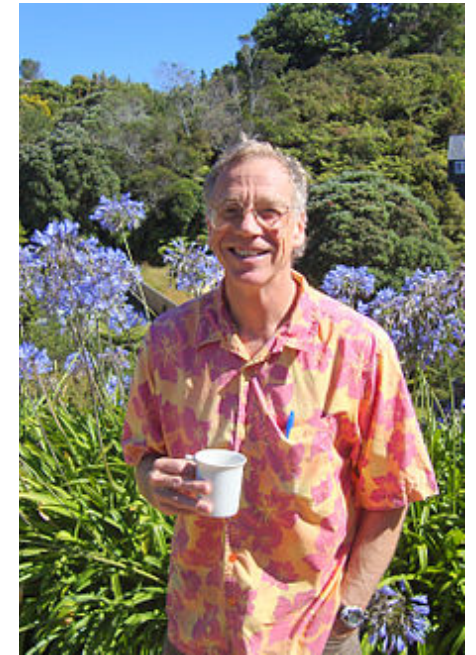
Quantum Computing with Noninteracting Bosons

Scott Aaronson (MIT)

Based on joint work with Alex Arkhipov

www.scottaaronson.com/papers/optics.pdf

This talk will involve **two** topics in which Mike Freedman played a pioneering role...



“Quantum computing beyond qubits”:

TQFT, nonabelian anyons...

- Yields new links between complexity and physics
- Can provide new implementation proposals

Quantum computing and #P:

Quantum computers can additively estimate the Jones polynomial, which is #P-complete to compute exactly

The Extended Church-Turing Thesis (ECT)

Everything feasibly
computable in the physical
world is feasibly computable
by a (probabilistic) Turing machine

But building a QC able to factor $n \gg 15$ is **damn** hard!

Can't CS "meet physics halfway" on this one?

i.e., show computational hardness in more easily-accessible quantum systems?

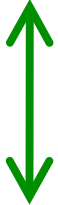
Also, factoring is an extremely "special" problem

Our Starting Point

$$\text{Det}(A) = \sum_{\sigma \in S_n} (-1)^{\text{sgn}(\sigma)} \prod_{i=1}^n a_{i, \sigma(i)}$$

$$\text{Per}(A) = \sum_{\sigma \in S_n} \prod_{i=1}^n a_{i, \sigma(i)}$$

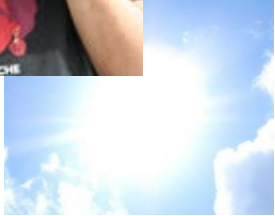
All I can say is, the bosons got the harder job



FERMIONS



e [Valiant]



BOSONS

This Talk: The Bosons Indeed Got The Harder Job

Valiant 2001, Terhal-DiVincenzo 2002, “folklore”:

A QC built of noninteracting fermions can be efficiently simulated by a classical computer

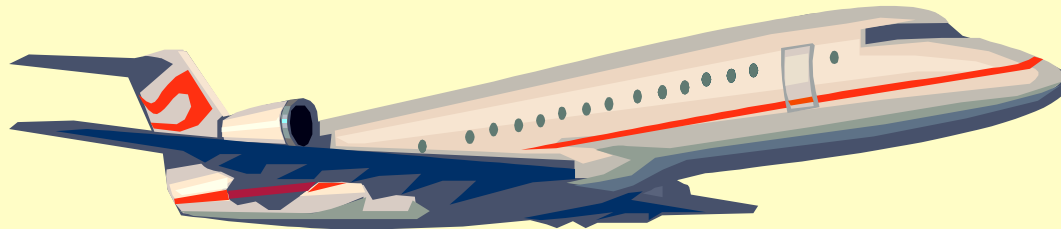
Our Result: By contrast, a QC built of noninteracting **bosons** can solve a sampling problem that’s hard for classical computers, under plausible assumptions

The Sampling Problem: Output a matrix $A \sim N(0,1)_C^{n \times n}$ with probability weighted by $|\text{Per}(A)|^2$

But wait!

If n-boson amplitudes correspond to nxn permanents,

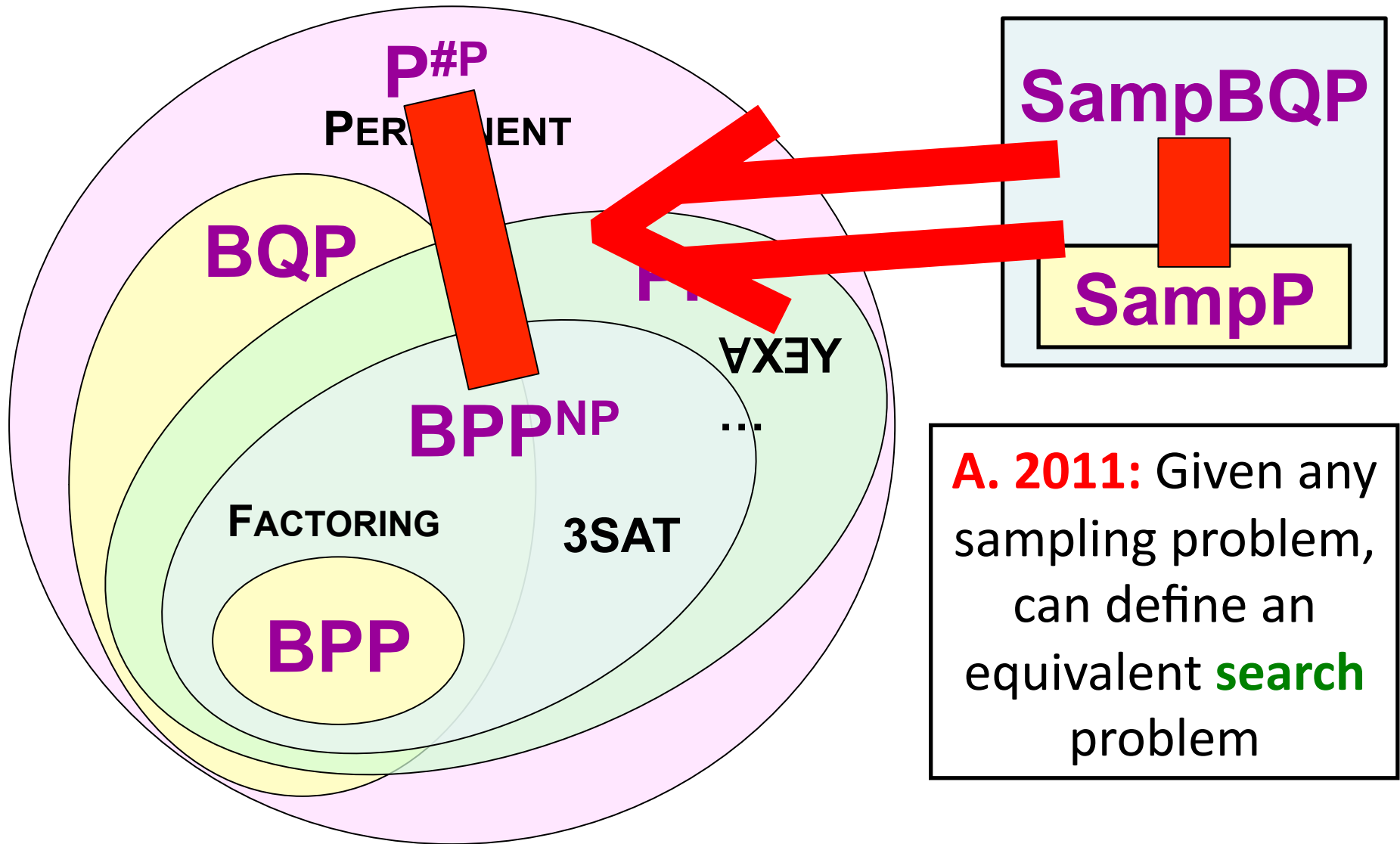
New result (from my flight here): Poly-time randomized algorithm to estimate the probability of **any** final state of a “boson computer,” to within $\pm 1/\text{poly}(n)$ additive error



Yes, but only up to additive error $\pm \frac{\|A\|}{\text{poly}(n)}$

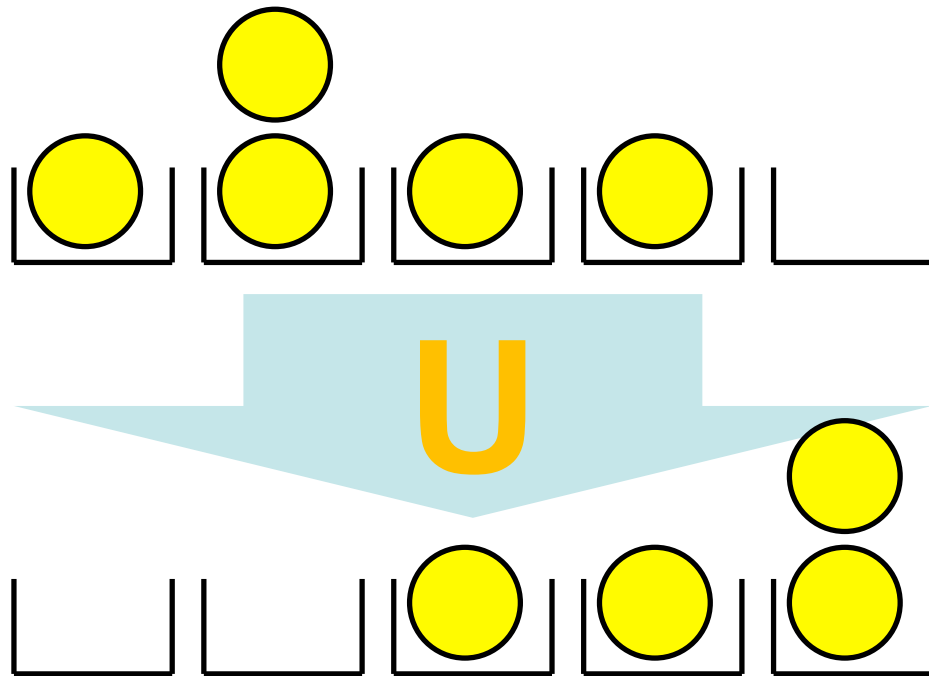
And Gurvits gave a poly-time classical randomized algorithm that estimates $|\text{Per}(A)|^2$ just as well!

Crucial step we take: switching attention to **sampling** problems



A. 2011: Given any sampling problem, can define an equivalent **search** problem

The Computational Model



Basis states: $|S\rangle = |s_1, \dots, s_m\rangle$,
 $s_i = \#$ of bosons in i^{th} mode
 $(s_1 + \dots + s_m = n)$

Standard initial state: $|1\rangle = |1, \dots, 1, 0, \dots, 0\rangle$

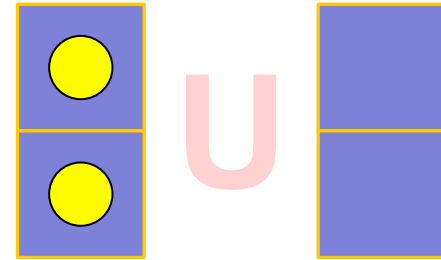
You get to apply any $m \times m$ mode-mixing unitary U

U induces a unitary $\varphi(U)$ on the n -boson states, whose entries are permanents of submatrices of U :

$$\langle S | \varphi(U) | T \rangle = \frac{\text{Per}(U_{S,T})}{\sqrt{s_1! \cdots s_m! t_1! \cdots t_m!}}$$

Example: The Hong-Ou-Mandel Dip

Suppose $U = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$.



Then $\Pr[\text{the two photons land in different modes}]$ is

$$|\text{Per}(U)|^2 = 0$$

$\Pr[\text{they both land in the first mode}]$ is

$$\left| \frac{1}{\sqrt{2}!} \text{Per} \left(\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \right) \right|^2 = \frac{1}{2}$$

For Card-Carrying Physicists

Our model corresponds to **linear optics**, with single-photon Fock-state inputs and nonadaptive photon-number measurements.

Physicists we consulted: “Sounds hard! But not as hard as building a universal QC”

Basic of the Hong-Ou-Mandel dip, where $n =$ big as possible

Our main results strongly suggest that such a generalized

Remark: No point in scaling this experiment much beyond 20 or 30 photons, since then a classical computer can't even verify the answers!

Ex

- Reliable single-photon sources
- Reliable photodetector arrays
- Getting a large n -photon coincidence probability

OK, so why is it hard to sample the distribution over photon numbers classically?

Given **any** matrix $A \in \mathbb{C}^{n \times n}$, we can construct an $m \times m$ unitary U (where $m \geq 2n$) as follows:

$$U = \begin{pmatrix} \varepsilon A & B \\ C & D \end{pmatrix}$$

Suppose we start with $|I\rangle = |1, \dots, 1, 0, \dots, 0\rangle$ (one photon in each of the first n modes), apply U , and measure.

Then the probability of observing $|I\rangle$ again is

$$p := \left| \langle I | \varphi(U) | I \rangle \right|^2 = \varepsilon^{2n} |\text{Per}(A)|^2$$

Claim 1: p is $\#P$ -complete to estimate (up to a constant factor)

Idea: Valiant proved that the PERMANENT is $\#P$ -complete.

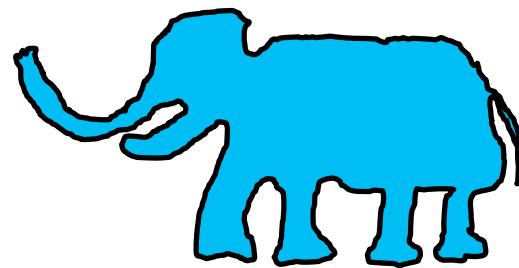
Can use a classical reduction to go from a multiplicative approximation of $|\text{Per}(A)|^2$ to $\text{Per}(A)$ itself.

Claim 2: Suppose we had a fast classical algorithm for linear-optics sampling. Then we could estimate p in BPP^{NP}

Idea: Let M be our classical sampling algorithm, and let r be its randomness. Use approximate counting to estimate $\Pr[M(r)\text{outputs } |I\rangle]$

Conclusion: Suppose we had a fast classical algorithm for linear-optics sampling. Then $\text{P}^{\#P} = \text{BPP}^{\text{NP}}$.

The Elephant in the Room



Our whole result hinged on the difficulty of estimating a **single, exponentially-small probability** p —but what about noise and error?

The “right” question: can a classical computer efficiently sample a distribution with **$1/\text{poly}(n)$ variation distance** from the linear-optical distribution?

Our Main Result: Suppose it can. Then there’s a **BPP^{NP}** algorithm to estimate $|\text{Per}(A)|^2$, with high probability over a Gaussian matrix $A \sim N(0,1)_{\mathbb{C}}^{n \times n}$

Our Main Conjecture

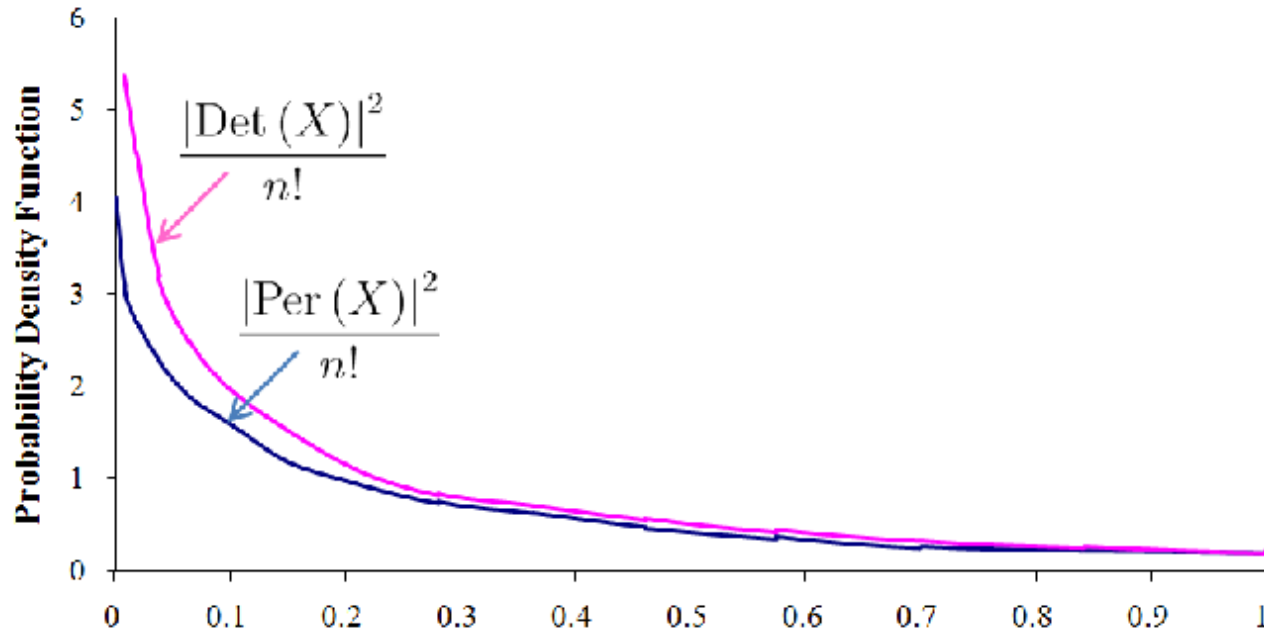
Estimating $|\text{Per}(A)|^2$, for most Gaussian matrices A , is a **#P**-hard problem

If the conjecture holds then even a **noisy** n -photon

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First step: Understand the **distribution** of $|\text{Per}(A)|^2$ for Gaussian A

Related Result: The KLM Theorem

Theorem (Knill, Laflamme, Milburn 2001): Linear optics with **adaptive measurements** can do universal QC

Yields an alternate proof of our first result (fast **exact** classical algorithm $\Rightarrow \mathbf{P}^{\#\mathbf{P}} = \mathbf{BPP}^{\mathbf{NP}}$)

A., last month: KLM also yields an alternate proof of Valiant's Theorem, that the permanent is **#P**-complete!

To me, more "intuitive" than Valiant's original proof

Similarly, **Kuperberg 2009** used Freedman-Kitaev-Larsen-Wang to reprove the **#P**-hardness of the Jones polynomial

Open Problems

Prove our main conjecture (\$1,000)!

Can our model solve classically-intractable **decision** problems?

Similar hardness results for other quantum systems (besides noninteracting bosons)?

Bremner, Jozsa, Shepherd 2010: QC with commuting Hamiltonians can sample hard distributions

Fault-tolerance within the noninteracting-boson model?