“Mind the gap”
Solving optimization problems with a quantum computer
A.P. Young

http://physics.ucsc.edu/~peter

Talk at the London Centre for Nanotechnology, October 17, 2012

Quantum computers promise to accelerate some kinds of calculations in a remarkable manner. But as in present-day classical computing, hardware is only half the story: efficiency requires development of appropriate algorithms, such as the fast Fourier transform.

To apply a quantum computer to a broad class of problems, general-purpose algorithms are needed. One such method is the quantum adiabatic algorithm, in which the problem to be solved is coded into a Hamiltonian $H$. One prepares the quantum computer in the ground state of a reference Hamiltonian $H_R$ and then has it evolve under a time-dependent Hamiltonian $H(t)$ that gradually switches from $H_R$ to $H$. If the evolution is slow enough (“adiabatic”) the system ends up in the ground state of $H$, which contains information about the desired solution.

In a paper in *Physical Review E*, Itay Hen and Peter Young of the University of California, Santa Cruz, show that “slow enough” may be very slow indeed. The reason is that the time required for adiabatic evolution depends inversely on the gap in energies between the ground and first excited states of $H(t)$. Using computer simulations, Hen and Young show that for three classes of logic problems, the scaling of the gap is such that the computational time can be expected to grow exponentially with the size of the problem. The authors suggest that it might be possible to optimize the evolution of $H(t)$ to avoid the bottleneck associated with a vanishing gap. – *Ron Dickman*
Quantum computers promise to accelerate some kinds of calculations in a remarkable manner. But as in present-day classical computing, hardware is only half the story: efficiency requires development of appropriate algorithms, such as the fast Fourier transform.

To apply a quantum computer to a broad class of problems, general-purpose algorithms are needed. One such method is the quantum adiabatic algorithm, in which the problem to be solved is encoded into a Hamiltonian $H$. One prepares the quantum computer in the ground state of a reference time-dependent Hamiltonian $H(t)$ that gradually switches from $H_0$ to $H$. If the evolution is slow enough, the final state of $H$, which contains information about the desired solution.

In a paper, Itay Hen and A. P. Young of the University of California, Santa Cruz, show that “slow enough” may be very slow indeed. The authors state that for three classes of logic problems, the scaling of the gap is such that the computational time can be exponential. The authors suggest that it might be possible to optimize the evolution of $H(t)$ to avoid the bottle-neck on Dickman.
Synopsis: Mind the Gap

Quantum computers promise to accelerate some kinds of problems, but only half the story: efficiency requires development of effective algorithms, in which the problem to be solved is encoded in an appropriate Hamiltonian. In a paper, Itay Hen and A. P. Young show how the quantum adiabatic algorithm can exhibit exponential complexity for certain satisfiability problems. Using a reduction from the bipartite version of the non-Boolean satisfiability problem $\text{3SAT}$, they demonstrate that to find a satisfying truth assignment for a random 3-CNF formula, one must follow the state of a quantum system through a region of exponentially large measure.

Exploiting the quantum adiabatic theorem, they establish a lower bound on the time required to implement the algorithm. They show that this time exhibits exponential growth as the size of the problem increases. This result complements the recent quantum money scheme proposed by Scott Aaronson and Alex Arkhipov, in which exponential complexity is required to distinguish between two types of quantum states.

The paper raises intriguing questions about the limits of quantum algorithms and the relative power of quantum and classical computation. It also highlights the importance of developing efficient quantum algorithms for solving real-world problems.
Plan
Plan

What is a quantum computer?
Plan

What is a quantum computer?

What could an eventual quantum computer do better than a classical computer? (Shor and Grover)
<table>
<thead>
<tr>
<th>Plan</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>What is a quantum computer?</strong></td>
</tr>
<tr>
<td><strong>What could an eventual quantum computer do better than a classical computer?</strong> (Shor and Grover)</td>
</tr>
<tr>
<td>Could a quantum computer be of <strong>broader utility</strong> by efficiently solving “<strong>optimization problems</strong>”, of broad interest in science and engineering (including industry)?</td>
</tr>
</tbody>
</table>
Plan

<table>
<thead>
<tr>
<th>What is a quantum computer?</th>
</tr>
</thead>
<tbody>
<tr>
<td>What could an eventual quantum computer do better than a classical computer? (Shor and Grover)</td>
</tr>
</tbody>
</table>

Could a quantum computer be of **broader utility** by efficiently solving **“optimization problems”**, of broad interest in science and engineering (including industry)?

- The **Quantum Adiabatic Algorithm (QAA)**
### Plan

<table>
<thead>
<tr>
<th>What is a quantum computer?</th>
</tr>
</thead>
<tbody>
<tr>
<td>What could an eventual quantum computer do better than a classical computer? (Shor and Grover)</td>
</tr>
<tr>
<td>Could a quantum computer be of <strong>broader utility</strong> by efficiently solving &quot;<strong>optimization problems</strong>&quot;, of broad interest in science and engineering (including industry)?</td>
</tr>
</tbody>
</table>

- The Quantum Adiabatic Algorithm (QAA)
- The Quantum Monte Carlo Method (QMC)
## Plan

<table>
<thead>
<tr>
<th>What is a quantum computer?</th>
</tr>
</thead>
<tbody>
<tr>
<td>What could an eventual quantum computer do better than a classical computer? (Shor and Grover)</td>
</tr>
<tr>
<td>Could a quantum computer be of <strong>broader utility</strong> by efficiently solving “<strong>optimization problems</strong>”, of broad interest in science and engineering (including industry)?</td>
</tr>
</tbody>
</table>

- The **Quantum Adiabatic Algorithm (QAA)**
- The **Quantum Monte Carlo Method (QMC)**
- Results for **SAT** problems
Plan

<table>
<thead>
<tr>
<th>What is a quantum computer?</th>
</tr>
</thead>
<tbody>
<tr>
<td>What could an eventual quantum computer do better than a classical computer? (Shor and Grover)</td>
</tr>
</tbody>
</table>

Could a quantum computer be of broader utility by efficiently solving “optimization problems”, of broad interest in science and engineering (including industry)?

- The Quantum Adiabatic Algorithm (QAA)
- The Quantum Monte Carlo Method (QMC)
- Results for SAT problems
- Comparison with a classical algorithm (WALKSAT)
# Plan

<table>
<thead>
<tr>
<th>What is a quantum computer?</th>
</tr>
</thead>
<tbody>
<tr>
<td>What could an eventual quantum computer do better than a classical computer? (Shor and Grover)</td>
</tr>
</tbody>
</table>

Could a quantum computer be of **broader utility** by efficiently solving “**optimization problems**”, of broad interest in science and engineering (including industry)?

- The **Quantum Adiabatic Algorithm** (QAA)
- The **Quantum Monte Carlo Method** (QMC)
- Results for **SAT** problems
- Comparison with a classical algorithm (**WALKSAT**)
- Results for a **spin glass** problem
Plan

What is a quantum computer?

What could an eventual quantum computer do better than a classical computer? (Shor and Grover)

Could a quantum computer be of broader utility by efficiently solving “optimization problems”, of broad interest in science and engineering (including industry)?

• The Quantum Adiabatic Algorithm (QAA)
• The Quantum Monte Carlo Method (QMC)
• Results for SAT problems
• Comparison with a classical algorithm (WALKSAT)
• Results for a spin glass problem
• Conclusions
Plan

What is a quantum computer?

What could an eventual quantum computer do better than a classical computer? (Shor and Grover)

Could a quantum computer be of broader utility by efficiently solving “optimization problems”, of broad interest in science and engineering (including industry)?

• The Quantum Adiabatic Algorithm (QAA)
• The Quantum Monte Carlo Method (QMC)
• Results for SAT problems
• Comparison with a classical algorithm (WALKSAT)
• Results for a spin glass problem
• Conclusions

Note: the QAA is inspired by physics and QMC is a technique from physics
What is a quantum computer?

Classical computer: bit is 0 or 1
Quantum computer: qubit: linear superposition of 0 and 1.

N qubits: linear superposition of $2^N$ basis states
Actions on the quantum state act on all $2^N$ basis states

⇒ Quantum Parallelism
What is a quantum computer?

Classical computer: bit is 0 or 1
Quantum computer: qubit: linear superposition of 0 and 1.

\( N \) qubits: linear superposition of \( 2^N \) basis states

Actions on the quantum state act on all \( 2^N \) basis states

\[ \Rightarrow \text{Quantum Parallelism} \]

But: to get the result, need to make a measurement

\[ \Rightarrow \text{Don’t get } 2^N \text{ results. Rather there are } 2^N \text{ probabilities and one gets one result} \text{ according to these probabilities.} \]
What is a quantum computer?

Classical computer: bit is 0 or 1
Quantum computer: qubit: linear superposition of 0 and 1.

N qubits: linear superposition of $2^N$ basis states
Actions on the quantum state act on all $2^N$ basis states

⇒ Quantum Parallelism

But: to get the result, need to make a measurement

⇒ Don’t get $2^N$ results. Rather there are $2^N$ probabilities and one gets one result according to these probabilities.

Seems that quantum mechanics is not useful for computing Nonetheless, in some cases useful results can be obtained by doing clever processing before the measurement.
What is a quantum computer?

**Classical computer:** bit is 0 or 1

**Quantum computer:** qubit: linear superposition of 0 and 1.

N qubits: linear superposition of $2^N$ basis states

Actions on the quantum state act on all $2^N$ basis states

$\Rightarrow$ Quantum Parallelism

**But:** to get the result, need to make a measurement

$\Rightarrow$ Don’t get $2^N$ results. Rather there are $2^N$ probabilities and one gets one result according to these probabilities.

**Seems** that quantum mechanics is not useful for computing

Nonetheless, in some cases useful results can be obtained by doing clever processing before the measurement.

The most famous is Shor’s algorithm for factoring integers
Integer factoring
Integer factoring

Let $N = p \cdot q$, where $p$ and $q$ are prime. Let $N$ have $n$ bits. How hard is it to factor $N$?

Classical, $\sim \exp(n^{1/3} \cdot (\log n)^{2/3})$ (exponential)
i.e hard. This is the basis of RSA encryption method

Quantum (Shor), $\sim n^3$ (polynomial)

Shor uses a “Quantum Fourier Transform”

Needs $3n$ qubits plus ???? for error correction.

Need coherence during the running of the algorithm.

$n \sim 1000$ would be useful
**Integer factoring**

Let \( N = p \times q \), where \( p \) and \( q \) are prime. Let \( N \) have \( n \) bits. How hard is it to factor \( N \)?

**Classical**, \( \sim \exp(n^{1/3} \times (\log n)^{2/3}) \) (exponential)

i.e hard. **This is the basis of RSA encryption method**

**Quantum (Shor)**, \( \sim n^3 \) (polynomial)

Shor uses a “Quantum Fourier Transform”

Needs 3n qubits plus ????? for error correction.

Need coherence during the running of the algorithm.

\( n \sim 1000 \) would be useful

\( \Rightarrow \) We are very far from having a useful quantum computer
Integer factoring

Let $N = p \cdot q$, where $p$ and $q$ are prime. Let $N$ have $n$ bits. How hard is it to factor $N$?

Classical, $\sim \exp(n^{1/3} \cdot (\log n)^{2/3})$ (exponential)

i.e hard.

Quantum (Shor), $\sim n^3$ (polynomial)

Shor uses a “Quantum Fourier Transform”

Needs $3n$ qubits plus ????? for error correction.

Need coherence during the running of the algorithm.

$n \sim 1000$ would be useful

⇒ We are very far from having a useful quantum computer

Still: interesting to investigate what could be done with a quantum computer if and when one will eventually be built.
RSA Encryption

Alice wants to send a message to Bob down a public channel.

Bob sends to Alice a “public key” \( N \), a product of 2 large random primes \( p \) and \( q \), i.e. \( N = p \cdot q \), and an encoding integer \( e \) which has no factors in common with \( (p-1)(q-1) \) (e.g. \( N \) has 1024 bits).

Alice’s message \( m \) is a binary string (\( m < N \)). She forms the encoded message \( m' \) from

\[
m' = m^e \mod N
\]

Alice sends \( m' \) to Bob down the public channel.

Bob knows his “private key” \( d \), the decoding integer, which is determined by

\[
d \cdot e = 1 \mod (p-1)(q-1)
\]

which is easily found (generalized Euclid) if one knows \( p \) and \( q \).

Bob computes \( (m')^d \mod N \) which is Alice’s message:

\[
m = (m')^d \mod N
\]
Optimization problems

Shor’s algorithm (and Grover’s, searching an unstructured data base of size $N$ with $\sim \sqrt{N}$ operations rather than $\sim N/2$) is rather specialized.

Would a quantum computer also be useful for more general problems, such as optimization problems, i.e. minimizing a function of $N$ variables with constraints?

Of interest in many fields in science and engineering.

Here we will take “Problem Hamiltonians” (i.e. the function to be minimized) which involve binary variables, 0 or 1, (or equivalently Ising spins $\sigma^z = \pm 1$).

How could we try to solve such optimization problems on a quantum computer?

An idea from physics ....
Quantum Adiabatic Algorithm

Proposed by Farhi et. al (2001) to solve hard optimization problems on a quantum computer.

\[ H(t) = [1 - s(t)]H_D + s(t)H_P \]

\( H_D \) (g.s.) adiabatic? \( H_P \) (g.s.?)

\( H_P \) is the problem Hamiltonian, depends on the \( \sigma_i^z \)
\( H_D \) is the driver Hamiltonian \( = -h \sum \sigma_i^x \)

\( 0 \leq s(t) \leq 1, \quad s(0) = 0, \quad s(T) = 1 \)

\( T \) is the running time

The quantum computer simulates \( H(t) \). System starts in ground state of driver Hamiltonian. If process is adiabatic (and \( T \to 0 \)), it ends in g.s. of problem Hamiltonian, and problem is solved. Minimum \( T \) is the “complexity”.
Quantum Adiabatic Algorithm

Proposed by Farhi et. al (2001) to solve hard optimization problems on a quantum computer.

\[ \mathcal{H}(t) = [1 - s(t)] H_D + s(t) H_P \]

\( H_D \) (g.s.) \hspace{1cm} \text{adiabatic?} \hspace{1cm} \mathcal{H}_P \) (g.s.?)

\( H_P \) is the problem Hamiltonian, depends on the \( \sigma^z_i \)
\( H_D \) is the driver Hamiltonian \( = -h \sum \sigma^x_i \)

\( 0 \leq s(t) \leq 1, \quad s(0) = 0, \quad s(T) = 1 \)
\( T \) is the running time

The quantum computer simulates \( H(t) \). System starts in ground state of driver Hamiltonian. If process is adiabatic (and \( T \to 0 \)), it ends in g.s. of problem Hamiltonian, and problem is solved.
Minimum \( T \) is the "complexity".
Quantum Adiabatic Algorithm

Proposed by Farhi et. al (2001) to solve hard optimization problems on a quantum computer.

\[ \mathcal{H}(t) = [1 - s(t)]\mathcal{H}_D + s(t)\mathcal{H}_P \]

\( \mathcal{H}_D \) (g.s.) \quad \text{adiabatic?} \quad \mathcal{H}_P \) (g.s.?)

\( \mathcal{H}_P \) is the problem Hamiltonian, depends on the \( \sigma_i^z \)
\( \mathcal{H}_D \) is the driver Hamiltonian \( = -h \sum_i \sigma_i^x \)

\[ 0 \leq s(t) \leq 1, \quad s(0) = 0, \quad s(T) = 1 \]

\( T \) is the running time

The quantum computer simulates \( \mathcal{H}(t) \). System starts in ground state of driver Hamiltonian. If process is adiabatic (and \( T \rightarrow 0 \)), it ends in g.s. of problem Hamiltonian, and problem is solved.

Minimum \( T \) is the “complexity”.

Is \( T \) exponential or polynomial in the problem size \( N \)?
Early Numerics

Early numerics, Farhi et al. for very small sizes $N \leq 20$, on a particular problem found the time varied only as $N^2$, i.e. polynomial!

But possible “crossover” to exponential at larger sizes?

To explore large sizes, need techniques from statistical physics, Quantum Monte Carlo.
Quantum Phase Transition

Bottleneck is likely to be a quantum phase transition (QPT) where the gap to the first excited state is very small.
Bottleneck is likely to be a quantum phase transition (QPT) where the gap to the first excited state is very small.
Quantum Phase Transition

Bottleneck is likely to be a quantum phase transition (QPT) where the gap to the first excited state is very small.
Bottleneck is likely to be a quantum phase transition (QPT) where the gap to the first excited state is very small.
Bottleneck is likely to be a quantum phase transition (QPT) where the gap to the first excited state is very small.

Landau Zener Theory:
To stay in the ground state the time needed is proportional to $\Delta E_{\text{min}}^{-2}$.
Bottleneck is likely to be a quantum phase transition (QPT) where the gap to the first excited state is very small.

Landau Zener Theory:
To stay in the ground state the time needed is proportional to $\Delta E_{\text{min}}^{-2}$.

Using QMC, we compute $\Delta E$ for different $s$: $\rightarrow \Delta E_{\text{min}}$
Quantum Monte Carlo

We do a sampling of the $2^N$ states (so statistical errors).

Study equilibrium properties of a quantum system by simulating a classical model with an extra dimension, imaginary time, $\tau$, where $0 \leq \tau < 1/T$.

Not perfect, but the only numerical method available for large $N$.

We use the “stochastic series expansion” method for Quantum Monte Carlo simulations which was pioneered by Anders Sandvik.

$$ Z \equiv \text{Tr} e^{-\beta \mathcal{H}} = \sum_{n=0}^{\infty} \frac{\text{Tr} \left( -\beta \mathcal{H} \right)^n}{n!} $$

Stochastically sum the terms in the series.
Examples of results with the SSE code

Time dependent correlation functions decay with $\tau$ as a sum of exponentials

$$\langle A(\tau)A(0)\rangle - \langle A \rangle^2 = \sum_{n \neq 0} |\langle 0|A|n\rangle|^2 \exp[-(E_n - E_0)\tau]$$

For large $\tau$ only first excited state contributes, $\rightarrow$ pure exponential decay

Small size, $N = 24$, excellent agreement with diagonalization.

Large size, $N = 128$, good quality data, slope of straight line $\rightarrow$ gap.
Dependence of gap on $s$

Results for the dependence of the gap to the first excited state, $\Delta E$, with $s$, for one instance of 1-in-3 SAT with $N = 64$. The gap has a minimum for $s$ about 0.66 which is the bottleneck for the QAA.
Dependence of gap on $s$

Results for the dependence of the gap to the first excited state, $\Delta E$, with $s$, for one instance of 1-in-3 SAT with $N = 64$.

The gap has a minimum for $s$ about 0.66 which is the bottleneck for the QAA.

Mind this gap
Dependence of gap on s

Results for the dependence of the gap to the first excited state, $\Delta E$, with $s$, for one instance of 1-in-3 SAT with $N = 64$.

The gap has a minimum for $s$ about 0.66 which is the bottleneck for the QAA.

We compute the minimum gap for many (50) instances for each size $N$ and look how the median minimum gap varies with size.
In satisfiability problems (SAT) we ask whether there is an assignment of $N$ bits which satisfies all of $M$ logical conditions ("clauses"). We assign an energy to each clause such that it is zero if the clause is satisfied and a positive value if not satisfied.

i.e. We need to determine if the ground state energy is 0.

We take the ratio of $M/N$ to be at the satisfiability threshold, and study instances with a "unique satisfying assignment" (USA).

(so gap to 1st excited state has a minimum whose value indicates the complexity.)

These SAT problems are "NP-complete", a category of hard problems for which the time is exponential with classical algorithms, at least in the worst case.
“Locked” 1-in-3 SAT

The clause is formed from 3 bits picked at random. The clause is satisfied (has energy 0) if one is 1 and the other two are 0 (in terms of spins one is -1 (green) and the other two are +1 (red)). Otherwise it is not satisfied (the energy is 1).

Example of a satisfying assignment with N=7, M = 5.

\[ \mathcal{H}_P = \sum_{\text{clauses}} \left( \frac{\sigma_1^z + \sigma_2^z + \sigma_3^z - 1}{2} \right)^2 \]

(V. Choi)
**Locked 1-in-3**

Plots of the median minimum gap (average over 50 instances)

Clearly the behavior of the minimum gap is exponential.
Comparison with a classical algorithm, WalkSAT: I

WalkSAT is a classical, heuristic, local search algorithm. It is a reasonable classical algorithm to compare with QAA. We have compared the running time of the QAA for the three SAT problems studied with that of WalkSAT.

For QAA, Landau-Zener theory states that the time is proportional to \( 1/(\Delta E_{\text{min}})^2 \) (neglecting \( N \) dependence of matrix elements).

For WalkSAT the running time is proportional to number of “bit flips”.

We write the running time as proportional to \( \exp(\mu N) \).

We will compare the values of \( \mu \) among the different models and between QAA and WalkSAT.
Comparison with a classical algorithm, WalkSAT: II

Exponential behavior for both QAA and WalkSAT

The trend is the same in both QAA and WalkSAT. 3-XORSAT is the hardest, and locked 1-in-3 SAT the easiest.
Comparison with a classical algorithm, WalkSAT: III

<table>
<thead>
<tr>
<th>Model</th>
<th>QAA</th>
<th>WalkSAT</th>
<th>Ratio</th>
</tr>
</thead>
<tbody>
<tr>
<td>1-in-3</td>
<td>0.084(3)</td>
<td>0.0505(5)</td>
<td>1.66</td>
</tr>
<tr>
<td>2-in-4</td>
<td>0.126(7)</td>
<td>0.0858(8)</td>
<td>1.47</td>
</tr>
<tr>
<td>3-XORSAT</td>
<td>0.159(2)</td>
<td>0.1198(4)</td>
<td>1.32</td>
</tr>
</tbody>
</table>

These results used the simplest implementation of the QAA for instances with a USA.

Exponential complexity in both cases. QAA not better than WalkSAT.

Values of $\mu$ (where time $\sim \exp[\mu N]$).
A “spin glass” on a random graph:

For simplicity we put the spins in a regular random graph, each site having exactly three neighbor (3-regular). Spins prefer to be antiparallel, an antiferromagnet (but see next slide)

The problem Hamiltonian is

\[ H_P = \frac{1}{2} \sum_{\langle i,j \rangle} \left( 1 + \sigma_i \sigma_j \right) \]

Note the symmetry under

\[ \sigma_i \rightarrow -\sigma_i, \; \forall i \]

“Replica” theory indicates that these 2-SAT-like problems are different from K-SAT problems for K > 2. (Hence we study it here.)

Note: there are large loops
Spin Glass on a random graph: II

Cannot form an “up-down” antiferromagnet because of loops of odd length. In fact, it is a “spin glass”.

Adding the driver Hamiltonian there is a quantum phase transition at $s = s^*$ above which the symmetry is spontaneously broken. “Cavity” calculations (Gosset, Zamponi) find $s^* \approx 0.36$

Investigated the problem near $s^*$ ($s \leq 0.5$). Also just considered instances with a “unique satisfying assignment” (apart from the degenerate state related by flipping all the spins). (These are exponentially rare.)
Spin glass on a random graph: III

Farhi, Gosset, Hen, Sandvik, Shor, Young, Zamponi

Comparison with cavity theory. Expected to be accurate but not exact for \( s > s^* \) (spin glass phase). (Would need “full replica symmetry breaking” (RSB) which is too hard.) Some difference may be due to different ensembles (cavity calculation has random instances, QMC those with a USA).
Spin glass on a random graph: III (Gap)

For larger sizes, a fraction of instances have **two minima**, one fairly close to \( s^* \) (\( \approx 0.36 \)) and other at larger \( s \) in the spin glass phase.

Figure shows an example for \( N = 128 \).

Hence did 2 analyses

(i) **Global** minimum in range (up to \( s=0.5 \))

(ii) If two minima, just take the **local** minimum near \( s^* \).

\[ \Delta E_{\text{min}} \sim \begin{cases} 0.29 \exp(-0.014 N) & \text{Global} \\ 0.26 \exp(-0.011 N) & \text{Local} \end{cases} \]

\[ E_{\text{min}} \sim \begin{cases} 5.31 N^{-0.96} & \text{Global} \\ 2.79 N^{0.78} & \text{Local} \end{cases} \]
Spin Glass: IV (Summary)

For this spin glass the QAA succeeds at the quantum critical point (polynomial gap).

However, it appears not to succeed at larger values of $s$ in the spin glass phase (exponential fit preferred over power-law for large sizes). But:

• This depends crucially on the last point ($N = 160$)

• A stretched exponential $\exp(-c N^x)$ ($x < 1$) also works pretty well, e.g. $x = 1/2$ (figure). If this is the correct answer we would say that the QAA does succeed.
Conclusions
Conclusions

• Simple application of QAA seems to fail:
Conclusions

- Simple application of QAA seems to fail:
  - At the (first order transition) for SAT problems
Conclusions

- Simple application of QAA seems to fail:
  - At the (first order transition) for SAT problems
  - In the spin glass phase, beyond the transition, in the spin glass model
Conclusions

• Simple application of QAA seems to fail:
  • At the (first order transition) for SAT problems
  • In the spin glass phase, beyond the transition, in the spin glass model
• This is just the simplest implementation of the QAA. Perhaps if one could find clever paths in Hamiltonian space one could better. A subject for future research.
Conclusions

• Simple application of QAA seems to fail:
  • At the (first order transition) for SAT problems
  • In the spin glass phase, beyond the transition, in the spin glass model

• This is just the simplest implementation of the QAA. Perhaps if one could find clever paths in Hamiltonian space one could better. A subject for future research.

Thank You