Optimisation via Adiabatic Quantum Computing

Vivek Katial

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About Me

- Vivek Katial (vkatial@student.unimelb.edu.au)
 - PhD Candidate (Optimisation on Quantum Computers)

Talk Structure

- Background
- Proposed Research
- Overview of Literature
- Current Progress
- What's next?

Main Research Question

 What instances characteristics of optimisation problems make them predisposed to being solved on a Quantum Computer?

Background

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• $|\psi(t)\rangle$ is our state vector, H(t) is the time dependent Hamiltonian. A Hamiltonian of an n-qubit system H(t) is given by $2^n \times 2^n$ matrix.

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 Loosely speaking, the adiabatic theorem tells us that if we vary from H_B to H_P slowly enough the system will remain in its ground state. This fact is a direct result of the Adiabatic Theorem [1].

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• Where g(t) is the difference between the first two smallest eigenvalues of H(t)

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- The basic SAT formulation can be described as follows: Given a boolean formula (AND \land , OR \lor , NOT \neg) over n variables $(z_1, z_2, ..., z_n)$. Can one set z_i 's in a manner such that the Boolean formula is true?

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- A clause is an expression which the variables must satisfy. For example $z_1 \wedge z_2 \implies z_1 = z_2 = 1$

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• However, our satisfying assignment is only $\vec{z} = 1111$

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- We then evolve our system with the following interpolation:

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• Finally, complete measurement and our solution is our final state $|\psi(t=T)\rangle$

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 - 2. Entropy of Entanglement

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- Ultimately, we can build a prediction model which maps instance features of different 3SAT instances to their inherent "Quantumness"

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- This methodology can then be further extended to other optmisation problems that can be mapped to Quadratic Unconstrained Binary Optimisation (QUBO) problems.

Algorithm Selection

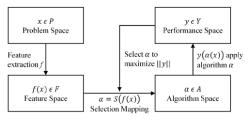
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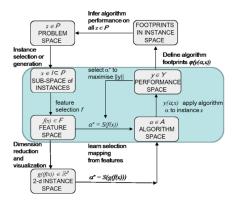


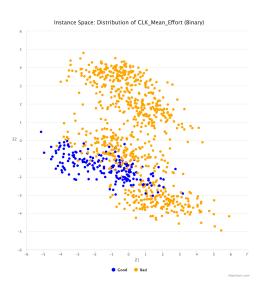
 However, what we are interested in is probing the strengths and weaknesses of AQC for different instances of SAT.

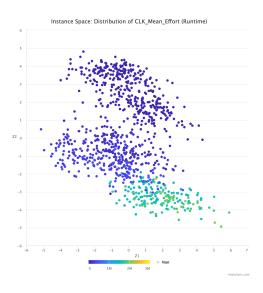
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- 6. Balance Features
- 7. DPLL Probing Features
- 8. Local Search Probing Features
- 9. LP-Based Features

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- Their research indicated that as $N \to \infty$ the system is expected to lead to an exponentially small gap, and hence an exponential complexity.
- Farhi et al. [13] also investigated different evolution paths and their results suggested it is possible to overcome the exponentially small minimum gap by selecting random initial Hamiltonians

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- However, their results also showed that for large *N*, the minimum gap scaling is exponential and can be associated with a *quantum phase transition* [14].
- Hauke et al. [15] ran simulations of adiabatic quantum optimization with n=16. Their results indicated that large entanglement entropy has little significance for the success probability of the optimization task.

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- Gabor et al. [17] shows that the phase transition from 3SAT persists in some form (but possibly to a lesser extent) in AQC via real experimental results on D-WAVE.
- Most signficant for us is that none of this work approaches instances in a robust manner such as the MATILDA framework

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- Developed working environment, set up GitHub, scripts, organizational tools enabling an efficient working environment.

4. Developed an architecture to run reproducible experiments at scale

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- 5. Provisioned all infrastructure required for above on Melbourne Research Cloud and SPARTAN

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- Developed a working implementation of Adiabatic Quantum Computing on different instances of 3SAT.

Next Steps (untill Confirmation)

1. Complete Literature Review of current advances in AQC.

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- 2. Implement scripts to generate **hard** instances of 3SAT with respect to *instance characteristics*.

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- 2. Implement scripts to generate **hard** instances of 3SAT with respect to *instance characteristics*.
- Apply Instance Space Methodology from MATILDA to find decision boundaries for Entanglement Entropy and Minimum Energy Gap

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- 8. Extend research to look at other optimisation problems

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