Optimisation via Adiabatic Quantum Computing

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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.

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

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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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- The 3-SAT problem in particular is a problem where each clause is comprised of 3 literals.

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

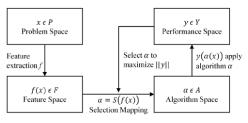
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 However, what we are interested in is probing the strengths and weaknesses of AQC for different instances of SAT.

Instance Space Methodology

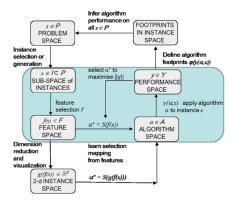
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- 6. Balance Features
- 7. DPLL Probing Features
- 8. Local Search Probing Features
- 9. LP-Based Features

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- 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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- 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.

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- 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

About Me

- Vivek Katial (vkatial@student.unimelb.edu.au)
 - PhD Candidate (Optimisation on Quantum Computers)
- Check out the slides at https://tinyurl.com/vkatial-preconfirmation

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