### Optimisation via Quantum Computing

Vivek Katial

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#### Introduction

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#### About Me

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

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#### Talk Structure

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

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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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#### Adiabatic Quantum Computing (AQC)

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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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# Adiabatic Quantum Computing (AQC)

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

• To conduct the computation we evolve  $|\psi(t)\rangle$  till time t=T such that  $|\psi(t=T)\rangle$  encodes the answer. The computation is done using a Hamiltonian which linearly interpolates between  $H_B$  and  $H_P$ . Specifically as below:

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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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## Mapping 3SAT to AQC

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

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- To do this we have simulated AQC and probe the "quantumness" of each instance:
  - 1. Minimum Energy Gap  $g_{\min}$
  - 2. Entropy of Entanglement

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- To survey the instances we will deploy the instance space analysis methodology [4–6]
- Ultimately, we can build a prediction model which maps instance features of different 3SAT instances to their inherent "Quantumness"

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 Given that we have access to IBM's Quantum Computer, we can also use this approach on QAOA and produce results from a universal quantum computer.

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- Given that we have access to IBM's Quantum Computer, we can also use this approach on QAOA and produce results from a universal quantum computer.
- This methodology can then be further extended to other optimisation problems that can be mapped to Quadratic Unconstrained Binary Optimisation (QUBO) problems.

## Algorithm Selection

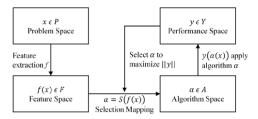
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 However, what we are interested in is probing the strengths and weaknesses of AQC for different instances of SAT.

## Instance Space Methodology

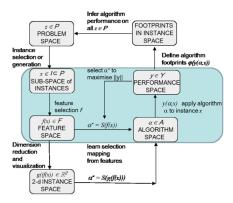
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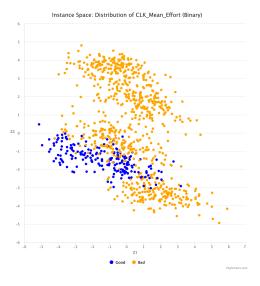
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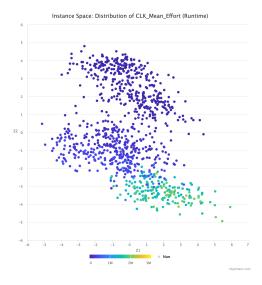
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#### Hardness Features for 3SAT

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

### Quantum Computing Research into Hard SAT problems

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- However in 2009, Young et al [12]. demonstrated via monte-carlo simulations of N=256 that some USA instances have a quantum phase transition.
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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

### Quantum Computing Research into Hard SAT problems

• Latorre et al. probed the entropy of entanglement for 250 USA Instances of Exact-Cover [14], their results showed that entropy of entanglement scales linearly with *N*.

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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 optimisation with n=16. Their results indicated that large entanglement entropy has little significance for the success probability of the optimisation 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

## **Current Progress**

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- Conducting literature review on current advances in Adiabatic Quantum Computing.
- Developed working environment, set up GitHub, scripts, organizational tools enabling an efficient working environment.

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- Provisioned all infrastructure required for above on Melbourne Research Cloud and SPARTAN
- 6. 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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# Next Steps (untill Confirmation)

Extend implementation to work on QAOA and VQE with QisKit

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- 5. Learn C++ effectively, this will be required when implementing larger simulations.
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- 7. Further advance knowledge of Quantum Physics (possibly through taking a course in Quantum Computing)
- 8. Extend research to look at other optimisation problems

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