

# Quadratization in Pseudo-Boolean Optimization and Adiabatic Quantum Computing

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A collaborative, evolving, open review paper on  $k$ -local to 2-local transformations (quadratizations) in classical computing, quantum annealing, and universal adiabatic quantum computing.

# Part I

## Diagonal Hamiltonians (pseudo-Boolean functions)

### I. METHODS THAT INTRODUCE ZERO AUXILIARY VARIABLES

#### A. Deduction Reduction (Deduc-reduc)

##### Summary

We look for *deductions* (e.g.  $b_1b_2 = 0$ ) that hold at the global minimum. These can be found by *a priori* knowledge of the given problem, or by enumerating solutions of a small subset of the variables. We can then substitute high-order terms using the low-order terms of the deduction, and add on a penalty term to preserve the ground states [13].

##### Cost

- 0 auxiliary variables needed.
- The computational cost of the search for deductions is difficult to estimate. The approximate worst-case complexity is  $\mathcal{O}(n^{d+1}2^m)$  where  $m$  is the number of variables in a ‘small’ problem,  $n$  is the total number of qubits and  $d$  is the maximum degree of deductions we are searching for. We suggest  $10 \leq m \leq 20$ , so that a small problem involves checking roughly 1,000 to 1,000,000 states, and  $d = 2$ . See the appendix for more details.

##### Pros

- No auxiliary variables needed.

##### Cons

- When deductions cannot be determined naturally (as in the Ramsey number determination problem, see Example IX), deductions need to be found by ‘brute force’, which scales exponentially with respect to  $m$ . For highly connected systems (systems with a large number of non-zero  $\alpha_{ij}$  coefficients), the value of  $m$  required to find even one deduction can be prohibitively large.

##### Example

Consider the Hamiltonian:

$$H_{4\text{-local}} = b_1b_2(10 + b_3 + b_3b_4) + b_1(b_3 - 3) + b_2(b_3 - 2b_3 - b_4 - 1) + F(x_3, x_4, x_5, \dots, x_N) \quad (1)$$

where  $F$  is any polynomial in  $b_i$  for  $i \geq 3$ .

Since the 10 coefficient of  $b_1b_2$  is greater than the sum of all of the other coefficients involving  $b_1$  or  $b_2$ , it must be the case that  $b_1b_2 = 0$ . Specifically, for the 4 assignments of  $(b_3, b_4)$ , we see that  $b_1b_2 = 0$  at every minimum of  $H_{4\text{-local}} - F$ .

Using deduc-reduc we have:

$$H_{2\text{-local}} = 12b_1b_2 + b_1(b_3 - 3) + b_2(b_3 - 2b_3 - b_4 - 1) + F(x_3, x_4, x_5, \dots, x_N) \quad (2)$$

which has the same global minima as  $H_{4\text{-local}}$  but one fewer quartic and one fewer cubic term.

##### Bibliography

- Original paper, with application to integer factorization: [13].

## B. ELC Reduction

### *Summary*

An Excludable Local Configuration (ELC) is a partial assignment of variables that make it impossible to achieve the minimum. We can therefore add a term that corresponds to the energy of this ELC without changing the solution to the minimization problem. In practice we can eliminate all monomials with a variable in which a variable is set to 0, and reduce any variable set to 1. Given a general Hamiltonian we can try to find ELCs by enumerating solutions of a small subset of variables in the problem [6].

### *Cost*

- 0 auxiliary variables needed.
- For  $n$  qubits there is  $\binom{n}{m}$  ways to choose  $m$  of them and  $2^m$  assignments of these  $m$  variables, therefore  $\mathcal{O}\left(2^m \binom{n}{m}\right)$  operations required to enumerate all possible cases for all possible subsets of size  $m \leq n$  variables.
- Approximate methods exist which have been shown to be much faster and give good approximations to the global minimum [6].

### *Pros*

- No auxiliary variables needed.

### *Cons*

- No known way to find ELCs except by 'brute force', which scales exponentially with respect to  $m$ .
- ELCs do not always exist.

### *Example*

Consider the Hamiltonian:

$$H_{3\text{-local}} = b_1 b_2 + b_2 b_3 + b_3 b_4 - 4b_1 b_2 b_3. \quad (3)$$

If  $b_1 b_2 b_3 = 0$ , no assignment of our variables will we be able to reach a lower energy than if  $b_1 b_2 b_3 = 1$ . Hence this gives us 12 ELCs, and one example is  $(b_1, b_2, b_3) = (1, 0, 0)$  which we can use to form the polynomial:

$$H_{2\text{-local}} = H_{3\text{-local}} + 4b_1(1 - b_2)(1 - b_3) \quad (4)$$

$$= b_1 b_2 + b_2 b_3 + b_3 b_4 + 4b_1 - 4b_1 b_2 - 4b_1 b_3. \quad (5)$$

In both cases Eqs. (3) and (5), the only global minima occur when  $b_1 b_2 b_3 = 1$ .

### *Bibliography*

- Original paper and application to computerized image denoising: [6].

### C. Groebner Bases

#### *Summary*

Given a set of polynomials, a Groebner basis is another set of polynomials that have exactly the same zeros. The advantage of a Groebner basis is it has nicer algebraic properties than the original equations, in particular they tend to have smaller degree polynomials. The algorithms for calculating Groebner bases are generalizations of Euclid's algorithm for the polynomial greatest common divisor.

Work has been done in the field of 'Boolean Groebner bases', but while the variables are Boolean the coefficients of the functions are in  $\mathbb{F}_2$  rather than  $\mathbb{Q}$ .

#### *Cost*

- 0 auxiliary variables needed.
- $\mathcal{O}(2^{2^n})$  in general,  $\mathcal{O}(d^{n^2})$  if the zeros of the equations form a set of discrete points, where  $d$  is the degree of the polynomial and  $n$  is the number of variables [14].

#### *Pros*

- No auxiliary variables needed.
- General method, which can be used for other rings, fields or types of variables.

#### *Cons*

- Best algorithms for finding Groebner bases scale double exponentially in  $n$ .
- Only works for Hamiltonians whose minimization corresponds to solving systems of discrete equations (RICHARD, why is this the ONLY case?).

#### *Example*

Consider the following pair of equations:

$$b_1 b_2 b_3 b_4 + b_1 b_3 + b_2 b_4 - b_3 = b_1 + b_1 b_2 + b_3 - 2 = 0. \quad (6)$$

Feeding these to Mathematica's `GroebnerBasis` function, along with the binarizing  $b_1(b_1 - 1) = \dots = b_4(b_4 - 1) = 0$  constraints, gives a Groebner basis:

$$\{b_4 b_3 - b_4, b_2 + b_3 - 1, b_1 - 1\}. \quad (7)$$

From this we can immediately read off the solutions  $b_1 = 1$ ,  $b_2 = 1 - b_3$  and reduce the problem to  $b_3 b_4 - b_4 = 0$ . Solving this gives a final solution set of:  $(b_1, b_2, b_3, b_4) = (1, 0, 1, 0), (1, 0, 1, 1), (1, 1, 0, 0)$ .

#### *Bibliography*

- Reduction and embedding of factorizations of all bi-primes up to 200,000: [15].

## D. Split Reduction

### *Summary*

It has been shown in [16] that, if multiple runs of a minimization algorithm is permitted, it is possible to reduce a lot of the problem by conditioning on the most connected variables. We call each of these operations a *split*.

### *Cost*

Exponential in the number of splits, as the number of problems to solve doubles with every split.

### *Pros*

- This method can be applied to any problem and can be very effective on problems with a few very connected variables.

### *Cons*

- Exponential cost in the worst case.

### *Example*

Consider the simple objective function

$$H = 1 + b_1b_2b_5 + b_1b_6b_7b_8 + b_3b_4b_8 - b_1b_3b_4. \quad (8)$$

In order to quadratize  $H$ , we first have to choose a variable to split over. In this case  $b_1$  is the obvious choice since it is present in the most terms and contributes to the quartic term.

We then obtain two different problems:

$$H_0 = 1 + b_3b_4b_8 \quad (9)$$

$$H_1 = 1 + b_2b_5 + b_6b_7b_8 + b_3b_4b_8 - b_3b_4. \quad (10)$$

At this point, we could split  $H_0$  again and solve it entirely, or use a qubit we saved in the previous split to quadratize our only problem.

To solve  $H_1$ , we can split again on  $b_8$ , resulting in problems:

$$H_{1,0} = 1 + b_2b_5 + b_6b_7 \quad (11)$$

$$H_{1,1} = 1 + b_2b_5 + b_3b_4. \quad (12)$$

Now both of these problems are quadratic. Hence we have reduced our original, hard problem into 3 easy problems, requiring only 2 extra runs of our minimization algorithm, and without needing any auxiliary variables.

### *Bibliography*

- Original paper and application to Ramsey number calculation: [16].

## II. METHODS THAT INTRODUCE AUXILIARY VARIABLES TO QUADRATIZE A SINGLE TERM

### A. Negative Term Reduction

#### *Summary*

For a negative term  $-b_1 b_2 \dots b_k$ , introduce a single auxiliary variable  $b_a$  and make the substitution:

$$-b_1 b_2 \dots b_k = \min_{b_a} \left( (k-1 - \sum_i b_i) b_a \right). \quad (13)$$

#### *Cost*

- 1 auxiliary variable for each  $k$ -local term.

#### *Pros*

- All resulting quadratic terms are submodular (have negative coefficients).
- Can reduce arbitrary order terms with only 1 auxiliary.

#### *Cons*

- Only works for negative terms.

#### *Example*

$$H_{6\text{-local}} = -2b_1 b_2 b_3 b_4 b_5 b_6 + b_5 b_6, \quad (14)$$

has a unique minimum energy of -1 when all  $b_i = 1$ .

$$H_{2\text{-local}} = 2(5b_a - b_1 b_a - b_2 b_a - b_3 b_a - b_4 b_a - b_5 b_a - b_6 b_a) + b_5 b_6 \quad (15)$$

has the same unique minimum energy, and it occurs at the same place (all  $b_i = 1$ ), with  $b_a = 1$ .

#### *Bibliography*

- Original paper: [17].
- Discussion: [5], [7].

## B. Positive Term Reduction

### *Summary*

By considering the negated literals  $\bar{b}_i = 1 - b_i$ , we recursively apply the previous method to  $b_1 b_2 \dots b_k = -\bar{b}_1 b_2 \dots b_k + b_2 b_3 \dots b_k$ . The final identity is:

$$b_1 b_2 \dots b_k = \min_{b_a} \left( \sum_{i=1}^{k-2} b_{a_i} (k - i - 1 + b_i - \sum_{j=i+1}^k b_j) \right) + b_{k-1} b_k \quad (16)$$

### *Cost*

- $k - 2$  auxiliary variables for each  $k$ -local term.

### *Pros*

- Works for positive monomials.

### *Cons*

- $k - 1$  non-submodular quadratic terms.

### *Example*

$$b_1 b_2 b_3 b_4 = \min_{b_a} b_{a_1} (2 + b_1 - b_2 - b_3 - b_4) + b_{a_2} (1 + b_2 - b_3 - b_4) + b_3 b_4 \quad (17)$$

### *Bibliography*

- Summary: [\[18\]](#).

### C. Ishikawa's Symmetric Reduction (Positive Term Reduction)

#### *Summary*

This method rewrites a positive monomial using symmetric polynomials, so all possible quadratic terms are produced and they are all non-submodular:

$$b_1 \dots b_k = \min_{b_{a_1}, \dots, b_{a_{n_k}}} \left( \sum_{i=1}^{n_k} b_{a_i} \left( c_{i,d} \left( - \sum_{j=1}^k b_j + 2i \right) - 1 \right) + \sum_{i < j} b_i b_j \right) \quad (18)$$

where  $n_k = \lfloor \frac{k-1}{2} \rfloor$  and  $c_{i,k} = \begin{cases} 1, & i = n_d \text{ and } k \text{ is odd,} \\ 2, & \text{else.} \end{cases}$

#### *Cost*

- $\lfloor \frac{k-1}{2} \rfloor$  auxiliary variables for each  $k$ -order term
- $\mathcal{O}(kt)$  for a  $k$ -local Hamiltonian with  $t$  terms.

#### *Pros*

- Works for positive monomials. About half as many auxiliary variables for each  $k$ -order term as the previous method.

#### *Cons*

- $\mathcal{O}(k^2)$  quadratic terms are created, which may make chimerization more costly.
- $\frac{k(k-1)}{2}$  non-submodular terms.
- Worse than the previous method for quartics, with respect to submodularity.

#### *Example*

$$b_1 b_2 b_3 b_4 = \min_{b_a} (3 - 2b_1 - 2b_2 - 2b_3 - 2b_4) b_a + b_1 b_2 + b_1 b_3 + b_1 b_4 + b_2 b_3 + b_2 b_4 + b_3 b_4 \quad (19)$$

#### *Bibliography*

- Original paper and application to image denoising: [5].



## D. Asymmetric Reduction

### *Summary*

Similar to other methods of reducing one term, this method can reduce a positive cubic monomial into quadratic terms using only one auxiliary variable, while introducing fewer non-submodular terms than the symmetric version.

The identity is given by:

$$b_1 b_2 b_3 = \min_{b_a} (b_a - b_2 b_a - b_3 b_a + b_1 b_a + b_2 b_3) \quad (20)$$

$$= \min_{b_a} (b_a - b_1 b_a - b_3 b_a + b_2 b_a + b_1 b_3) \quad (21)$$

$$= \min_{b_a} (b_a - b_1 b_a - b_2 b_a + b_3 b_a + b_1 b_2) \quad (22)$$

$$(23)$$

### *Cost*

1 auxiliary qubit per positive cubic term.

### *Pros*

- Works on positive monomials.
- Fewer non-submodular terms than Ishikawa Reduction.

### *Cons*

- Only been shown to work for cubics.

### *Bibliography*

- Original paper and application to computer vision: [\[19\]](#).

## E. Bit flipping

### Summary

For any variable  $b$ , we can consider the negation  $\bar{b} = 1 - b$ . The process of exchanging  $b$  for  $\bar{b}$  is called *flipping*. Using bit-flipping, an arbitrary function in  $n$  variables can be represented using at most  $2^{(n-2)}(n-3) + 1$  variables, though this is a gross overestimate.

Can be used in many different ways:

1. Flipping positive terms and using II A, recursively;
2. For  $\alpha < 0$ , we can reduce  $\alpha \bar{b}_1 \bar{b}_2 \dots \bar{b}_k$  very efficiently to submodular form using II A. A generalized version exists for arbitrary combinations of flips in the monomial which makes reduction entirely submodular [5];
3. When we have quadratized we can minimize the number of non-submodular terms by flipping.
4. We can make use of both  $b_i$  and  $\bar{b}_i$  in the same Hamiltonian by adding on a sufficiently large penalty term:  $\lambda(b_i + \bar{b}_i - 1)^2 = \lambda(1 + 2b_i\bar{b}_i - b_i - \bar{b}_i)$ . This is similar to the ideas in reduction by substitution or deduc-reduc. In this way, given a quadratic in  $n$  variables we can make sure it only has at most  $n$  nonsubmodular terms if we are willing to use the extra  $n$  negation variables as well (so we have  $2n$  variables in total).

### Cost

- None, as replacing  $b_i$  with it's negation  $\bar{b}_i$  costs nothing except a trivial symbolic expansion.

### Pros

- Cheap and effective way of improving submodularity.
- Can be used to combine terms in clever ways, making other methods more efficient.

### Cons

- Unless the form of the Hamiltonian is known, spotting these 'factorizations' using negations is difficult.
- We need an auxiliary variable for each  $b_i$  for which we also want to use  $\bar{b}_i$  in the same Hamiltonian.

### Example

By bit-flipping  $b_2$  and  $b_4$ , i.e. substituting  $b_2 = 1 - \bar{b}_2$  and  $b_4 = 1 - \bar{b}_4$ , we see that:

$$H = 3b_1b_2 + b_2b_3 + 2b_1b_4 - 4b_2b_4 \quad (24)$$

$$= -3b_1\bar{b}_2 - \bar{b}_2b_3 - 2b_1\bar{b}_4 - \bar{b}_2\bar{b}_4 + 5b_1 + b_3 + 4\bar{b}_2 + 4\bar{b}_4 - 4. \quad (25)$$

The first expression is highly non-submodular while the second is entirely submodular.

### Bibliography

- Original paper: [5].

### III. METHODS THAT INTRODUCE AUXILIARY VARIABLES TO QUADRATIZE MULTIPLE TERMS WITH THE SAME AUXILIARIES

#### A. Reduction by Substitution

##### *Summary*

Pick a variable pair  $(b_i, b_j)$  and substitute  $b_i b_j$  with a new auxiliary variable  $b_{a_{i,j}}$ . Enforce equality in the ground states by adding some scalar multiple of  $b_i b_j - 2b_i b_{a_{i,j}} - 2b_j b_{a_{i,j}} + 3b_{a_{i,j}}$  or similar.

##### *Cost*

- 1 auxiliary variable per reduction.

##### *Pros*

- Variable can be used across the entire Hamiltonian, reducing many terms at once.
- Simple.

##### *Cons*

- Inefficient for single terms as it introduces many auxiliary variables compared to Ishikawa reduction, for example.
- Introduces quadratic terms with large positive coefficients, making them highly non-submodular.
- Determining optimal substitutions is expensive.

##### *Example*

We pick the pair  $(b_1, b_2)$  and combine.

$$b_1 b_2 b_3 + b_1 b_2 b_4 \mapsto b_3 b_a + b_4 b_a + b_1 b_2 - 2b_1 b_a - 2b_2 b_a + 3b_a \quad (26)$$

##### *Bibliography*

- Original paper: [20]

## B. FGBZ Reduction (Fix-Gruber-Boros-Zabih)

### Summary

Here we consider a set  $C$  of variables which occur in multiple monomials throughout the Hamiltonian. Each application 'rips out' this common component from each term [21][18].

Let  $\mathcal{H}$  be a set of monomials, where  $C \subseteq H$  for each  $H \in \mathcal{H}$  and each monomial  $H$  has a weight  $\alpha_H$ . The algorithm comes in 2 parts: when all  $\alpha_H > 0$  and when all  $\alpha_H < 0$ . Combining the 2 gives the final method:

1.  $\alpha_H > 0$

$$\sum_{H \in \mathcal{H}} \alpha_H \prod_{j \in H} b_j = \min_{b_a} \left( \sum_{H \in \mathcal{H}} \alpha_H \right) b_a \prod_{j \in C} b_j + \sum_{H \in \mathcal{H}} \alpha_H (1 - b_a) \prod_{j \in H \setminus C} b_j \quad (27)$$

2.  $\alpha_H < 0$

$$\sum_{H \in \mathcal{H}} \alpha_H \prod_{j \in H} b_j = \min_{b_a} \sum_{H \in \mathcal{H}} \alpha_H \left( 1 - \prod_{j \in C} b_j - \prod_{j \in H \setminus C} b_j \right) b_a \quad (28)$$

### Cost

- One auxiliary variable per application.
- In combination with II A, it can be used to make an algorithm which can reduce  $t$  positive monomials of degree  $d$  in  $n$  variables using  $n + t(d - 1)$  auxiliary variables in the worst case.

### Pros

- Can reduce the connectivity of a Hamiltonian, as it breaks interactions between variables.

### Cons

- $\alpha_H > 0$  method converts positive terms into negative ones of same order rather than reducing them, though these can then be reduced more easily.
- $\alpha_H < 0$  method only works for  $|C| > 1$ , and cannot quadratize cubic terms.

### Example

First let  $C = b_1$  and use the positive weight version:

$$b_1 b_2 b_3 + b_1 b_2 b_4 \mapsto 2b_{a_1} b_1 + (1 - b_{a_1}) b_2 b_3 + (1 - b_{a_1}) b_2 b_4 \quad (29)$$

$$= 2b_{a_1} b_1 + b_2 b_3 + b_2 b_4 - b_{a_1} b_2 b_3 - b_{a_1} b_2 b_4 \quad (30)$$

now we can use II A:

$$-b_{a_1} b_2 b_3 - b_{a_1} b_2 b_4 \mapsto 2b_{a_2} - b_{a_1} b_{a_2} - b_{a_2} b_2 - b_{a_2} b_3 + 2b_{a_2} - b_{a_1} b_{a_2} - b_{a_2} b_2 - b_{a_2} b_4 \quad (31)$$

$$= 4b_{a_2} - 2b_{a_1} b_{a_2} - 2b_{a_2} b_2 - b_{a_2} b_3 - b_{a_2} b_4. \quad (32)$$

### Bibliography

- Original paper and application to image denoising: [21].

## IV. METHODS THAT REPRODUCE THE FULL SPECTRUM

These techniques will transform  $k$ -local functions to 2-local functions that not only have the same ground state as the  $k$ -local function, but also the entire input/output spectrum of the  $k$ -local functions will be preserved in the low-lying energy space of the corresponding 2-local function (which also has higher energy states due to the auxiliary variables added).

### A. Recursive Order Reduction

#### *Summary*

For each  $k$ -local term an auxiliary variable and a quadratic penalty function to constrain the auxiliary variable, is used to reduce the term's order to  $(k - 1)$ -local. This is repeated recursively until the term is 2-local.

#### *Cost*

- $k - 2$  auxiliary variables to reduce each  $k$ -local terms.
- At most  $k t$  for a  $k$ -local Hamiltonian of  $t$  terms.

#### *Pros*

- Simple iterative method conducive to automated reduction.
- Chain like structure means that long range connectivity not required.

#### *Cons*

- Not very symmetric with respect to variables.
- Usually formulated in terms of  $b_i b_j$  interactions, therefore not conducive to specialized hardware design since hardware usually has native  $z_i z_j$  interactions.

#### *Example*

Let us assume we want to reproduce the spectrum of  $H_4 = b_1 b_2 b_3 b_4$ , first note that

$$P(b_{a_k}, b_i, b_j) = 3b_{a_k} + b_i b_j - 2b_i b_{a_k} - 2b_j b_{a_k} \quad (33)$$

will yield an energy of  $E = 0$  iff  $b_{a_k} = b_i b_j$ , and  $E > 0$  otherwise. We first reduce the order of the polynomial by 1 by applying this penalty term with a large positive prefactor  $\lambda$ .

$$H_{3\text{-local}} = b_1 b_2 b_{a_1} + \lambda(3b_{a_1} - 2b_{a_1}(b_3 + b_4) + b_3 b_4) \quad (34)$$

Performing the same trick again yields a 2-local Hamiltonian:

$$H_{2\text{-local}} = b_1 b_{a_2} + \lambda(3b_{a_2} - 2b_{a_2}(b_2 + b_{a_1}) + b_2 b_{a_1}) + \lambda(3b_{a_1} - 2b_{a_1}(b_3 + b_4) + b_3 b_4) \quad (35)$$

#### *Bibliography*

- Possibly the original paper [ask Boros and Gruber for advice on where it came first]: [24].
- Used in: [10, 23].

## B. Flag Based SAT Mapping

### *Summary*

This method is very similar to recursive order reduction, but uses gadgets to produce separate 3-SAT clauses which allow variables which ‘flag’ the state of pairs of other variables

### *Cost*

- Polynomial in size of Hamiltonian being simulated, but varies

### *Pros*

- Highly general and therefore conducive to proofs.

### *Cons*

- Designed for generality rather than efficiency.

### *Example*

To create a system which maps  $(b_1 = -1) \vee (b_2 = -1) \vee (b_3 = -1)$  (up to an energy shift), we use the following gadget:

$$H_{3\text{-SAT}} = - \sum_{i=1}^3 b_{a_i} - \frac{1}{2} \sum_{i=1}^3 b_i + \frac{1}{2} \sum_{i < j}^3 b_{a_i} b_{a_j} + \frac{1}{2} \sum_{i=1}^3 b_i b_{a_i}. \quad (36)$$

Implementing  $\lambda H_{3\text{-SAT}}$ , creates a situation where  $b_3$  is a ‘flag’ for  $b_1 = -1$  and  $b_2 = -1$  in other words  $b_3$  is constrained to be  $-1$  in the low energy manifold iff  $b_1 = -1$  and  $b_2 = -1$  and  $+1$  otherwise. It follows from the universality of 3 – SAT that these ‘flag’ clauses can be combined to map any spin Hamiltonian.

### *Bibliography*

- Paper showing the universality of the Ising spin models: [25].

### C. Chained Three Body Parity Operators

#### Summary

Goal: Guarantee that  $b_{a,k} = b_i b_j$  can be chained together to make large product terms consisting of  $b$ . The penalty term  $P(b_{a_k}, b_i, z_j) = \mp b_{a_k} b_i b_j$  guarantees that  $b_{a_k} = \pm b_i b_j$ . The three local term can be made from gadgets.

This method was originally only used to reproduce the ground state of high locality terms, but states of the "wrong" parity ( $q_k = \mp b_i b_j$ ) will all have the same energy as well, so it reproduces the full spectrum.

#### Cost

- The best known gadget for a three local Ising term uses one auxilliary qubit. Based on this gadget an  $n \geq 4$  body Ising term can be made using  $3 + 2(n - 4)$  auxilliary qubits.

#### Pros

- Natural transmon implementation [26].
- Chain like structure means that long range connectivity not required.

#### Cons

- Does not preserve degeneracy, ground state will retain original degeneracy, but excited states will have degeneracy multiplied by  $n - 3$
- Not very symmetric.

#### Example

Let us use this method to reproduce the spectrum of the following five local Hamiltonian  $b_1 b_2 b_3 b_4 b_5$ . The following three local gadget reproduces the spectrum of the three local term  $\pm b_1 b_2 b_3$

$$P_{\pm}(b_1, b_2, b_3; \lambda) = \lambda (b_1 b_2 + b_2 b_3 + b_3 b_1 + 2b_a(b_1 + b_2 + b_3)) \mp (b_1 + b_2 + b_3 + 2b_a). \quad (37)$$

Using three copies of this three local gadget as a building block, and using two additional auxilliary variables, the spectrum of the five local term  $b_1 b_2 b_3 b_4 b_5$  can be reproduced by the following Hamiltonian

$$H_{2\text{-local}} = P_+(b_1, b_2, b_{a_1}; \lambda) + P_+(b_{a_1}, b_3, b_{a_2}; \lambda) + P_+(b_{a_2}, b_4, b_5; \lambda). \quad (38)$$

#### Bibliography

- Original proposal with transmon implementation: [26].

## D. Symmetry Based Mappings

### Summary

Auxilliary qubits can be made to "count" the number of logical qubits in the 1 configuration. By applying single qubit terms to the auxilliary qubits, the spectrum of *any* permutation symmetric Hamiltonian can be reproduced.

### Cost

- For a  $k$  local coupler requires  $k$  auxilliary qubits.

### Pros

- Natural flux qubit implementation [32].
- Single gadget can reproduce any permutation symmetric spectrum.
- High degree of symmetry means this method is natural for some kinds of quantum simulations [33].

### Cons

- Requires coupling between all logical qubits and from all logical qubits to all auxilliary qubits.
- Requires single body terms of increasing strength as  $k$  is increased.

### Example

A 4 qubit gadget guarantees that the number of auxillary bits in the 0 state is equal to the number of logical bits in the 1 state

$$H_{4\text{-count}} = \sum_{i=2}^4 \sum_{j=1}^{i-1} z_i z_j - \frac{1}{2} \sum_{i=1}^4 z_i + \sum_{i=1}^4 \sum_{j=1}^4 z_i z_{a_j} + \frac{1}{2} (5z_{a_1} + z_{a_2} - 3z_{a_3} - 7z_{a_4}). \quad (39)$$

To replicate the spectrum of  $z_1 z_2 z_3 z_4$ , we add

$$H_{4\text{-local}} = z_{a_1} - z_{a_2} + z_{a_3} - z_{a_4} + \lambda H_{4\text{-count}}, \quad (40)$$

where  $\lambda$  is a large number.

For the spectrum of  $b_1 b_2 b_3 b_4$ , we implement,

$$H_{4\text{-local}} = -\frac{1}{2} z_{a_4} + \lambda H_{4\text{-count}}. \quad (41)$$

### Bibliography

- Paper on flux qubit implementation: [32]
- Paper on max-k-sat mapping: [34]
- Talk including use in quantum simulation: [33]



## Part II

### Hamiltonians quadratic in $z$ and linear in $x$ (Transverse Field Ising Hamiltonians)

The Ising Hamiltonian with a transverse field in the  $x$  direction is possible to implement in hardware:

$$H = \sum_i \left( \alpha_i^{(z)} z_i + \alpha_i^{(x)} x_i \right) + \sum_{ij} \left( \alpha_{ij}^{(zz)} z_i z_j \right). \quad (42)$$

#### E. ZZZ-TI-CBBK: Transverse Ising from ZZZ, by Cao, Babbush, Biamonte, and Kais (2015)

There is only one reduction in the literature for reducing a Hamiltonian term to the transverse Ising Hamiltonian, and it works on 3-local  $zzz$  terms, by introducing an auxiliary qubit with label  $a$ :

$$\begin{aligned} \alpha z_i z_j z_k &\rightarrow \alpha^I + \alpha_i^z z_i + \alpha_j^z z_j + \alpha_k^z z_k + \alpha_a^z z_a + \alpha_a^x x_a + \alpha_{ia}^{zz} z_i z_a + \alpha_{ja}^{zz} z_j z_a + \alpha_{ka}^{zz} z_k z_a \quad (43) \\ \alpha^I &= \frac{1}{2} \left( \Delta + \left( \frac{\alpha}{6} \right)^{2/5} \Delta^{3/5} \right) \\ \alpha_i^z &= -\frac{1}{2} \left( \left( \frac{7\alpha}{6} + \left( \frac{\alpha}{6} \right)^{3/5} \Delta^{2/5} \right) - \left( \frac{\alpha \Delta^4}{6} \right)^{1/5} \right) \\ \alpha_j^z &= \alpha_i^{(z)} \\ \alpha_k^z &= \alpha_i^{(z)} \\ \alpha_a^z &= \frac{1}{2} \left( \Delta - \left( \frac{\alpha}{6} \right)^{2/5} \Delta^{3/5} \right) \\ \alpha_a^x &= \left( \frac{\alpha \Delta^4}{6} \right)^{1/5} \\ \alpha_{ia}^{zz} &= -\frac{1}{2} \left( \left( \frac{7\alpha}{6} + \left( \frac{\alpha}{6} \right)^{3/5} \Delta^{2/5} \right) + \left( \frac{\alpha \Delta^4}{6} \right)^{1/5} \right) \\ \alpha_{ja}^{zz} &= \alpha_{ia}^{(zz)} \\ \alpha_{ka}^{zz} &= \alpha_{ja}^{(zz)} \end{aligned}$$

Including all coefficients and factorizing, we get:

$$\alpha z_i z_j z_k \rightarrow \left( \Delta + \frac{\alpha \Delta^4}{6} (z_i + z_j + z_k) \right) \left( \frac{1 - z_a}{2} \right) + \frac{\alpha \Delta^4}{6} x_a \quad (44)$$

$$+ \left( \left( \frac{\alpha}{6} \right)^{2/5} \Delta^{3/5} - \left( \frac{7\alpha}{6} + \left( \frac{\alpha}{6} \right)^{3/5} \Delta^{2/5} \right) (z_i + z_j + z_k) \right) \left( \frac{1 + z_a}{2} \right) \quad (45)$$

The low-lying spectrum (eigenvalues *and* eigenvectors) of the right side of Eq. (42) will match those of the left side to within a spectral error of  $\epsilon$  as long as  $\Delta = \mathcal{O}(\epsilon^{-5})$ .

#### Cost

- 1 auxiliary qubit
- 8 auxiliary terms not proportional to  $\mathbb{1}$ .

#### Example

**F. 1B1-TI: 1-by-1 gadget for  $ZZ \dots Z \rightarrow$  Transverse Ising (Present Work)**

***Cost***

- $k - 2$  auxiliary qubits
- auxiliary terms not proportional to  $\mathbb{1}$ .

***Example***

**G. SD-TI: Sub-division gadget for  $\mathbb{Z}\mathbb{Z} \dots \mathbb{Z} \rightarrow$  Transverse Ising (Present Work)**

*Cost*

- $k/2$

*Example*

## Part III

### General Quantum Hamiltonians

The most general time-independent Hamiltonian for  $n$  qubits acted on only by Pauli operators is:

$$H = \alpha I + \sum_i^n \left( \alpha_i^{(z)} z_i + \alpha_i^{(x)} x_i + \alpha_i^{(y)} y_i \right) + \quad (46)$$

$$\sum_{ij}^n \left( a_{ij}^{(zz)} z_i z_j + a_{ij}^{(zx)} z_i x_j + a_{ij}^{(zy)} z_i y_j + a_{ij}^{(xz)} x_i z_j + a_{ij}^{(xx)} x_i x_j + a_{ij}^{(xy)} x_i y_j + a_{ij}^{(yz)} y_i z_j + a_{ij}^{(yx)} y_i x_j + a_{ij}^{(yy)} y_i y_j \right) + \quad (47)$$

$$\sum_{ijk}^n \left( a_{ijk}^{(zzz)} z_i z_j z_k + a_{ijk}^{(zzx)} z_i z_j x_k + a_{ijk}^{(zzy)} z_i z_j y_k + a_{ijk}^{(xzz)} x_i z_j z_k + \dots + a_{ijk}^{(yyy)} y_i y_j y_k \right) + \quad (48)$$

$$\sum_{ijkl\dots n}^n \left( a_{ijkl\dots n}^{(zzzz\dots z)} z_i z_j z_k z_l \dots z_n + a_{ijkl\dots n}^{(zzzz\dots x)} z_i z_j z_k \dots x_n + a_{ijkl\dots n}^{(zzzz\dots y)} z_i z_j z_k \dots y_n + a_{ijkl\dots n}^{(zz\dots xz)} z_i z_j \dots x_{n-1} z_n + \dots + a_{ijkl\dots n}^{(yyy\dots y)} y_i y_j y_k \dots y_n \right) \quad (49)$$

We wish to find a 2-local Hamiltonian of the form of Eq. (??) or Eq. (??) that is equivalent to Eq. (45) in terms of reproducing certain desired properties (e.g. same ground state(s), same ground eigenvalue(s), same full eigenspectrum, etc.) to within the desired precision.

## V. 1-BY-1 GADGETS

A 1B1 gadget allows  $k$ -local terms to be quadratized one step at a time, where at each step the term's order is reduced by at most one. In each step, a  $k$ -local term is reduced to  $(k-1)$ -local, contrary to SD (sub-division) gadgets which can reduce  $k$ -local terms to  $(1/2)$ -local in one step.

### A. 1B1-KKR (Kempe, Kitaev, Regev, 2006)

#### Summary

Group all  $k$ -local terms together and express their sum as a sum of products of commuting matrices  $s_{ij}$ :

$$H_{k\text{-local}} = \sum_i \prod_j^k s_{ij} + H_{(k-1)\text{-local}}. \quad (50)$$

Define auxiliary qubits labeled by  $a_{ij}$  and make the transformation:

$$\sum_i \prod_j^k s_{ij} \mapsto \frac{\Delta}{4} \left( 9 - \sum_i \left( \sum_j z_{a_{ij}} \right)^2 - \sum_j z_{a_{ij}}^2 + \frac{1}{\sqrt[3]{6\Delta}} x_{a_{ij}} \right) + \Lambda. \quad (51)$$

The result will be a  $(k-1)$ -local Hamiltonian with the same low-lying spectrum as  $H_{k\text{-local}}$  to within  $\epsilon$  as long as  $\Delta = \Theta(\epsilon^{-3})$ .

#### Cost

- The number of auxiliary qubits needed is the number of  $s_{ij}$  matrices in Eq. (51).
- $\Delta = \Theta(\epsilon^{-3})$

## B. 1B1-OT (Oliviera & Terhal, 2008)

### *Summary*

We wish to reduce the  $k$ -local term:

$$H_{k\text{-local}} = \alpha \prod_j^k s_j. \quad (52)$$

Define one auxiliary qubit labeled by  $a$  and make the transformation:

$$H_{k\text{-local}} \rightarrow -\left(\frac{\alpha}{2}\right)^{1/3} \Delta^{2(1-r)} s_k \left(\frac{1-z_a}{2}\right) + \left(\frac{\alpha}{2}\right)^{1/3} \frac{\Delta^r}{\sqrt{2}} (s_{k-1} - s_{k-2}) x_a \quad (53)$$

$$+ \frac{1}{2} \left(\frac{\alpha}{2}\right)^{2/3} \left( \Delta^{r-1} s_{k-1} + \text{sgn}(\alpha) \sqrt{2} \Delta^{-1/4} \prod_j^{k-2} s_j \right)^2 + \frac{\alpha}{4} (1 + 2\text{sgn}^2 \alpha \Delta^{3/2-2r}) s_k. \quad (54)$$

The result will be a  $(k-1)$ -local Hamiltonian with the same low-lying spectrum as  $H_{k\text{-local}}$  to within  $\epsilon$  as long as  $\Delta = \Theta(\epsilon^{-3})$ .

### *Cost*

- Only 1 auxiliary qubit.
- $\Delta = \Theta(\epsilon^{-3})$

### *Example*

### *Bibliography*

**C. 1B1-CBBK (Cao, Babbush, Biamonte, Kais, 2015)**

Define one auxiliary qubit labeled by  $a$  and make the transformation:

$$H_{k\text{-local}} \rightarrow \left( \Delta + \left( \frac{\alpha}{2} \right)^{3/2} \Delta^{1/2} s_k \right) \left( \frac{1 - z_a}{2} \right) \quad (55)$$

$$- \frac{\alpha^{2/3}}{2} (1 + \text{sgn}^2 \alpha) \left( (2\alpha)^{2/3} \text{sgn}^2 \alpha + \alpha^{1/3} s_k - \sqrt[3]{2} \Delta^{1/2} \right) \left( \frac{1 + z_a}{2} \right) \quad (56)$$

$$+ \left( \frac{\alpha}{2} \right)^{1/3} \Delta^{3/4} \left( \prod_j^{k-2} s_j - \text{sgn}(\alpha) s_{k-1} \right) x_a + \text{sgn}(\alpha) \sqrt[3]{2} \alpha^{2/3} (\Delta^{1/2} + \Delta^{3/2}) \prod_j^{k-1} s_j. \quad (57)$$

The result is  $(k-1)$ -local and its low-lying spectrum is the same as that of  $H_{k\text{-local}}$  when  $\Delta$  is large enough.

## VI. SUBDIVISION GADGETS

Instead of recursively reducing  $k$ -local to  $(k-1)$ -local one reduction at a time, we can reduce  $k$ -local terms to  $(k/2)$ -local terms directly for even  $k$ , or to  $(k+1)/2$ -local terms directly for odd  $k$ . Since when  $k$  is odd we can add an identity operator to the  $k$ -local term to make it even, we will assume in the following that  $k$  is even, in order to avoid having to write floor and ceiling functions.

### A. SD-OT (Oliviera & Terhal, 2008)

#### *Summary*

For any  $k$ -local term, we can subdivide it into a product of two  $(k/2)$ -local terms, with an auxiliary qubit labeled  $a$ :

$$H_{k\text{-local}} \rightarrow \Delta \frac{1 - z_a}{2} + \sqrt{\frac{\Delta}{2}} \left( s_k - \prod_j^{k-1} s_j \right) x_a \quad (58)$$

#### *Cost*

- $\Delta = \frac{\|H_{(k-1)\text{-local}}\|^2}{2}$ .
- $\Delta = \Theta(\epsilon^{-2})$ .

#### *Pros*

- only one qubit to reduce  $k$  to  $\lceil k/2 \rceil + 1$

#### *Cons*

- Only beneficial for  $k \geq 5$ .

#### *Example*

#### *Bibliography*

## B. SD-CBBK (Cao, Babbush, Biamonte, Kais 2015)

### *Summary*

For any  $k$ -local term, we can subdivide it into a product of two  $(k/2)$ -local terms:

$$H_{k\text{-local}} = \alpha H_{1,(k/2)\text{-local}} H_{2,(k/2)\text{-local}} + H_{(k-1)\text{-local}}. \quad (59)$$

Define one qubit  $a$  and make the following Hamiltonian is  $(k/2)$ -local:

$$\Delta \frac{1 - z_a}{2} + |\alpha| \frac{1 + z_a}{2} + \sqrt{|\alpha| \Delta / 2} (\text{sgn}(\alpha) H_{1,(k/2)\text{-local}} - H_{2,(k/2)\text{-local}}) x_a \quad (60)$$

The result is a  $(k/2)$ -local Hamiltonian with the same low-lying spectrum as  $H_{k\text{-local}}$  for large enough  $\Delta$ . The disadvantage is that  $\Delta$  has to be larger. *Cost*

- $\Delta \geq \left( \frac{2|\alpha|}{\epsilon} + 1 \right) (|\alpha| + \epsilon + 22 \|H_{(k-1)\text{-local}}\|)$

### *Pros*

- only one qubit to reduce  $k$  to  $\lceil k/2 \rceil + 1$

### *Cons*

- Only beneficial for  $k \geq 5$ .

### *Example*

### *Bibliography*



**C. Duan et al (2011)**

*Summary*

*Cost*

*Example*

*Bibliography*

**D. BOA (Babbush, O’Gorman, Aspuru-Guzik, 2013)**

*Summary*

*Cost*

*Example*

*Bibliography*

## VII. DIRECT REDUCTION

Here we do not reduce  $k$  by one order at a time (1B1 reduction) or by  $k/2$  at a time (SD reduction), but we directly reduce  $k$ -local terms to 2-local terms.

### A. DR-JF (Jordan & Fahri)

*Summary*

*Cost*

*Example*

*Bibliography*

## VIII. NON-PERTURBATIVE EMBEDDINGS

### A. NP-SJ (Subasi & Jarzynski, 2016)

*Summary* Kato 1949.

*Cost*

*Example*

*Bibliography*

## Part IV

### Appendix

#### IX. FURTHER EXAMPLES

*Example* Here we show how deductions can arise naturally from the Ramsey number problem. Consider  $\mathcal{R}(4, 3)$  with  $N = 4$  nodes. Consider a Hamiltonian:

$$H = (1 - z_{12})(1 - z_{13})(1 - z_{23}) + \dots + (1 - z_{23})(1 - z_{24})(1 - z_{34}) + z_{12}z_{13}z_{14}z_{23}z_{24}z_{34}. \quad (61)$$

See [16] for full details of how we arrive at this Hamiltonian.

Since we are assuming we have no 3-independent sets, we know that  $(1 - z_{12})(1 - z_{13})(1 - z_{23}) = 0$ , so  $z_{12}z_{13}z_{23} = z_{12}z_{13} + z_{12}z_{23} + z_{13}z_{23} - z_{12} - z_{13} - z_{23} + 1$ . This will be our deduction.

Using deduc-reduc we can substitute this into our 6-local term to get:

$$H = 2(1 - z_{12})(1 - z_{13})(1 - z_{23}) + \dots + (1 - z_{23})(1 - z_{24})(1 - z_{34}) + \quad (62)$$

$$z_{14}z_{24}z_{34}(z_{12}z_{13} + z_{12}z_{23} + z_{13}z_{23} - z_{12} - z_{13} - z_{23} + 1). \quad (63)$$

We could repeat this process to remove all 5- and 4-local terms without adding any auxiliary qubits. Note in this case the error terms added by deduc-reduc already appear in our Hamiltonian.

#### ACKNOWLEDGMENTS

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