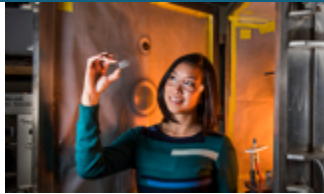
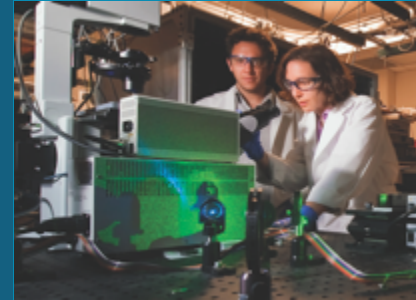




# Beating Random Assignment for Approximating Quantum 2-Local Hamiltonian Problems



*PRESENTED BY*

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# Quantum Bits Live in a Sphere



**Classical bit:  
(bit)**



OR



+1 = Head

-1 = Tail

State space

$\{+1, -1\}$

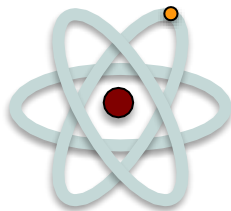
**Prob. bit:  
(p-bit)**



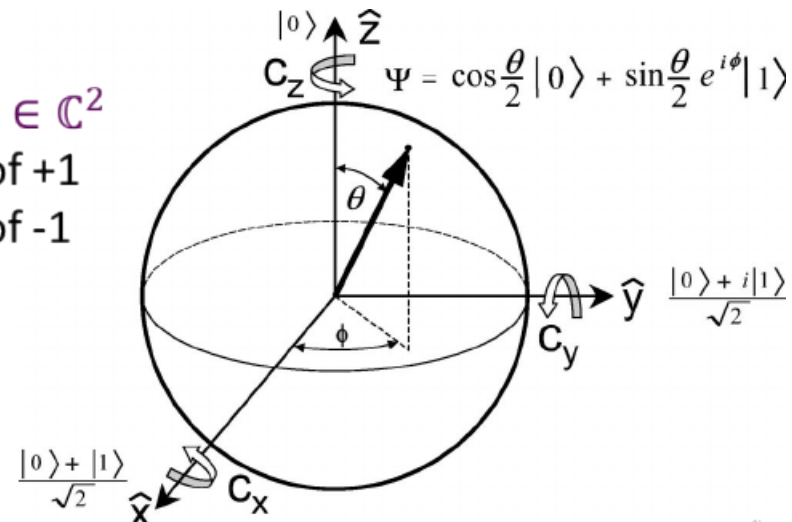
+1 with probability  $p$   
-1 with probability  $1-p$



**Quantum bit:  
(qubit)**



$\alpha|+1\rangle + \beta|-1\rangle \in \mathbb{C}^2$   
 $|\alpha|^2$ -probability of +1  
 $|\beta|^2$ -probability of -1



# How to describe a Generic distribution of qubits?



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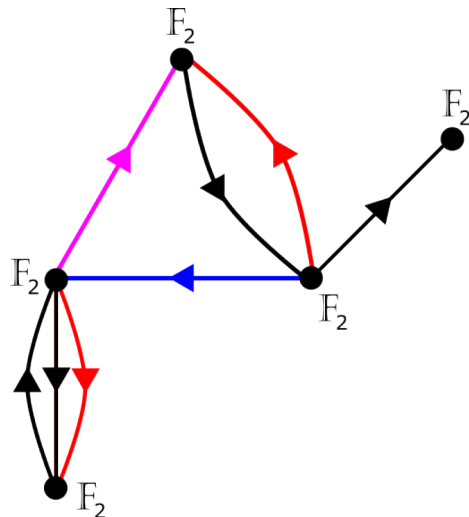
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# Classical and Quantum Combinatorial Optimization



## Classical



- Edges correspond to non-linear functions on vertices

$$f = \sum_{(j \rightarrow k) \in E} f_{jk}(v_j, v_k)$$

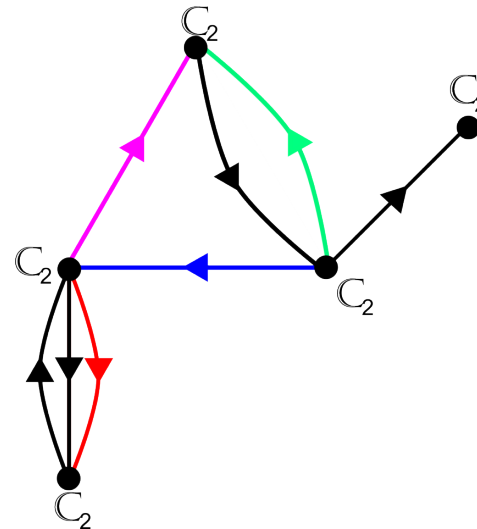
- Task is to find

$$\max_{\mu} \mathbb{E}_{\mu}[f]$$

- Example** Max Cut

$$f_{jk}(v_j, v_k) = v_j + v_k - 2v_j v_k$$

## Quantum (2-Local Hamiltonian Problem)



- Optimizing over tensor product

$$\mathbb{F}_2^{\oplus n} \leftrightarrow \mathbb{C}_2^{\otimes n}$$

- Edge functions now linear, obj is the same

$$\max_{\rho} \mathbb{E}_{\rho}[f] \approx \max_{\rho} \text{Tr} \left[ \sum_{(j \rightarrow k) \in E} h_{jk} \rho \right]$$

- $h_{jk}$  has “non-trivial” part only on  $j$  and  $k$ :

$$h_{12} = O_{12} \otimes \mathbb{I}_3 \otimes \cdots \otimes \mathbb{I}_n$$

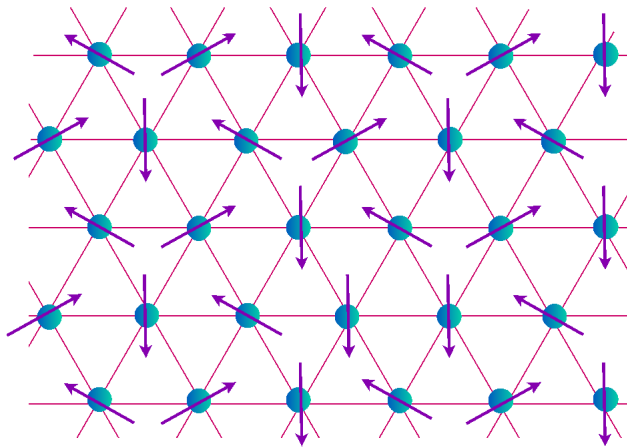
- Find largest e-value of exp. large matrix from a local description**





## Possible Energies for Physical systems

- Heisenberg model is fundamental for describing quantum magnetism, superconductivity, and charge density waves. We consider generalization of this problem



**Anti-ferromagnetic Heisenberg model:** roughly neighboring quantum particles aim to align in opposite directions. This kind of Hamiltonian appears, for example, as an effective Hamiltonian for so-called Mott insulators.  
[Image: Sachdev, arxiv:1203.4565]

## Complexity Theory

- 2-Local Hamiltonian problem is QMA complete -> **If we shouldn't expect to solve it, how well can we approximate it?**
- Field of Classical Approximation Algorithms very developed. **Very few results giving rigorous approx. algs. for quantum problems.**



# Approximation Algorithms and Ansätze



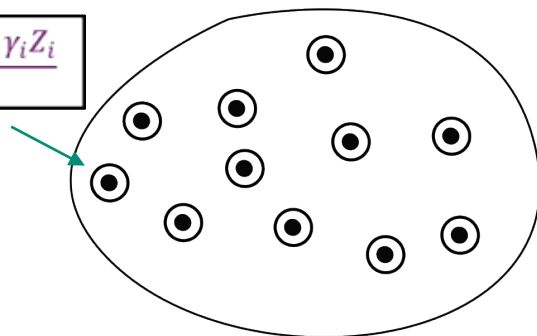
$$\frac{\text{Alg}(\{h_{ij}\})}{\lambda_{\max}(H)} \geq \alpha$$

Runs in poly time in  $n$ ,  
provable guarantee independent of instance

- Unlike classical combinatorial opt. not clear what “kind” of description is best
- **Ansatz**- “kind” of quantum state the algorithm outputs.
- **Example: Product State Ansatz**

$$\rho_i = \frac{I + \alpha_i X_i + \beta_i Y_i + \gamma_i Z_i}{2}$$

$$\rho = \prod \rho_i$$



- **Analogy**: Independent Coin Flips
- Overall distribution of independent variables can be specified with marginals





# Assumptions

Assumption	Reference
Graph-Dependent Assumption	[Bansal, Bravyi, Terhal, '07'], [Gharibian, Kempe, '12'], [Brandao, Harrow, '16], [Harrow, Montanaro, '17]
Quantum Max Cut	[Gharibian, Parekh, '19], [Anshu, Gosset, Morenz, '20] [PT '21] (x2)
Traceless	[Bravyi, Gosset, Koenig, Temme, '18]
Projectors	[Hallgren, Lee, P. '20] [This work]

- No “universal” approximation algorithm: generalizes max ind. set [Wocjan, Beth '03]
- Just like classical, people study 2-Local HP with assumptions
- **For us: 2-Local terms are projectors, i.e.**  $h_{12} = P_{12} \otimes \mathbb{I}_3 \otimes \cdots \otimes \mathbb{I}_n$ 
  - Assume further the rank of each  $h_{ij}$  is the same
  - 3 cases of interest corresponding to rank of  $P_{ij}$



# Our Contributions

**Thm** For **bipartite** traceless we get

$$\alpha = \frac{2 \ln(1 + \sqrt{2})}{3 \pi}$$

**Thm** If each  $h_{(i \rightarrow j)} = P_{\{i,j\}} \otimes \mathbb{I}_{[n] \setminus \{i,j\}}$  where  $\text{rank}(P) = k$ , there is an approx. alg. obtaining factor:

$$\alpha(k) = \begin{cases} 0.387 & \text{if } k = 1 \\ 0.585 & \text{if } k = 2 \\ 0.764 & \text{if } k = 3 \end{cases}$$

**Thm** If in addition each  $P$  is “strictly quadratic”, we obtain the following approx. factors:

$$\alpha(k) = \begin{cases} 0.467 & \text{if } k = 1 \\ 0.639 & \text{if } k = 2 \\ 0.805 & \text{if } k = 3 \end{cases}$$

Trivial (Random) Assignment:

$$\alpha(k) = \begin{cases} 1/4 & \text{if } k = 1 \\ 1/2 & \text{if } k = 2 \\ 3/4 & \text{if } k = 3 \end{cases}$$



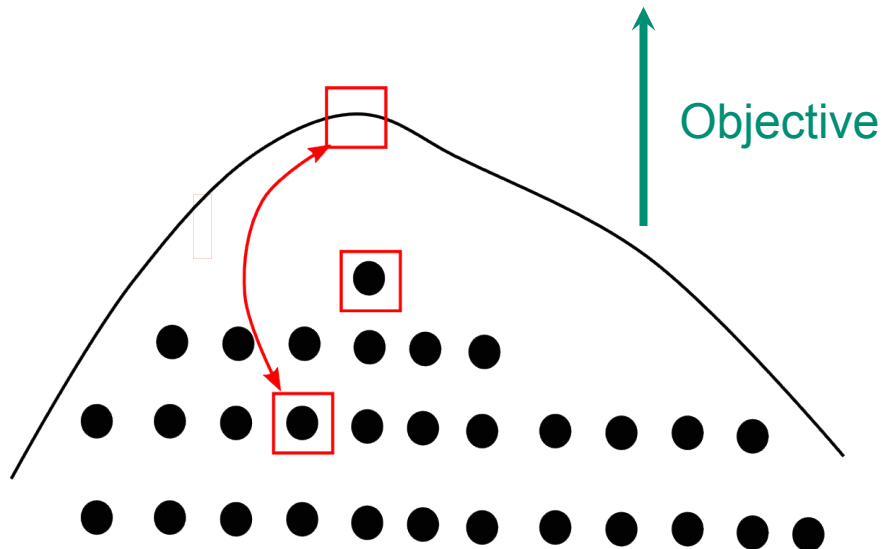
# How Does the Approx. Alg. Work?

## Classical Rounding-

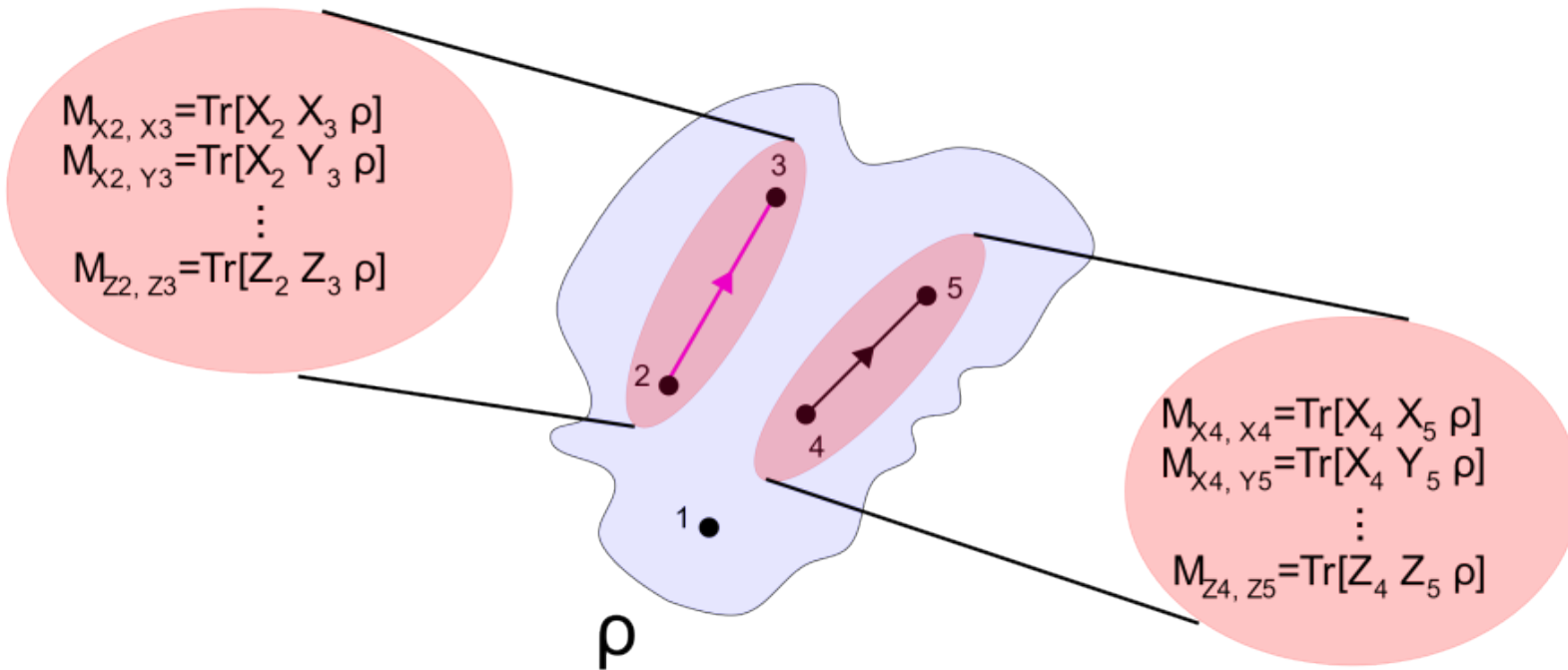
- Relax NP-hard optimization problem
- Use solution of relaxation to get element in  $\mathbb{F}_2^n$
- Analyze loss in obj. from rounding

## Quantum Rounding

- Problem is exp. large SDP
- Can we relax this to polynomially large SDP?



# Relaxation



- Define **Moment Matrix**  $M \in \mathbb{C}^{3n \times 3n}$

$$M_{\sigma j, \eta l} = \text{Tr}[\sigma_j \eta_l \rho] : \sigma, \eta \in \{X, Y, Z\}$$

- Pauli basis is complete, this description captures all **2-Local** statistics  $\Rightarrow$  Given **only** the moment matrix, we could calculate the objective



# The SDPs

- Hence, could equivalently optimize over  $M$ :

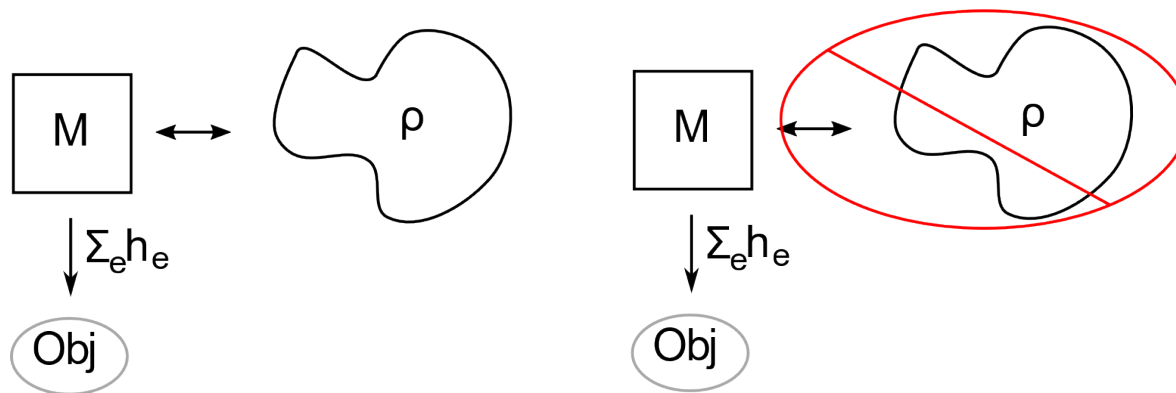
$$\begin{aligned} \max \quad & \sum_e \text{Tr}(C^e M) \\ \text{s.t.} \quad & M_{\sigma_j, \eta_l} = \text{Tr}(\sigma_j \eta_l \rho) \\ & \rho \in \mathcal{D}(2^n), \quad M \in \mathbb{R}^{3n \times 3n}, \succeq 0 \end{aligned}$$

$$v^T M v = \text{Tr}(T^2 \rho) \geq 0$$

For non-imaginary entries

- Haven't really done anything at this point**

- Equivalent to original SDP, merely represented 2-Local part redundantly
- Can “forget” correspondence to get poly-sized SDP!

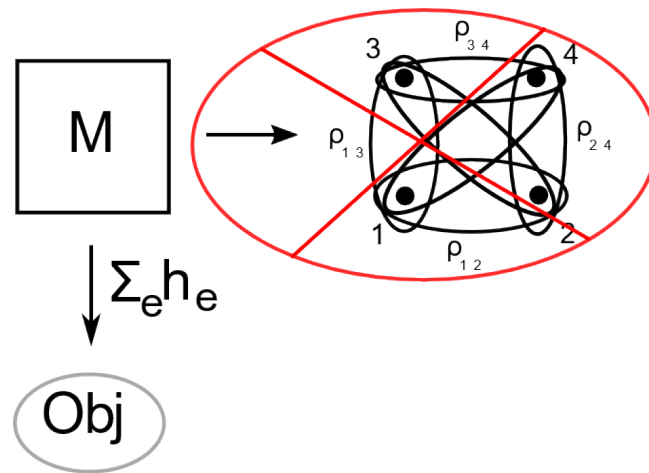




- **The catch?** In SDP, marginals will be globally inconsistent
- Otherwise, we would be solving a QMA-hard problem!
- $M$  may not even be consistent with a set of 2-local marginals
- SDP we study forces with property with additional variables
- Slight tightening of the 1<sup>st</sup> level of the quantum Lasserre Hierarchy

[Lasserre '01]

[Pironio, Navascués, Acín, '10]

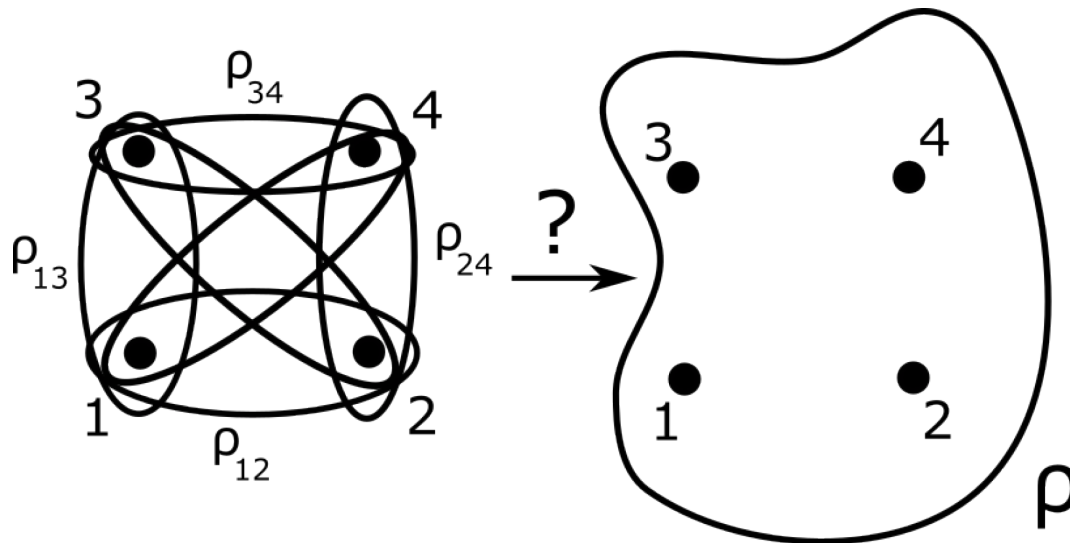


$$\begin{aligned}
 & \max \sum_e \text{Tr}(C^e M) \\
 & \text{s.t. } M_{\sigma_j, \eta_l} = \text{Tr}(\sigma_j \eta_l \rho) \forall \sigma, j, \eta, l \\
 & \quad M_{ij, ij} = 1, M_{ij, ik} = 0 \\
 & \quad \text{Tr}(\rho_{ij}) = 1 \forall i, j \\
 & \quad M \succeq 0, \rho_{ij} \succeq 0 \forall i, j \\
 & \quad M \in \mathbb{R}^{3n \times 3n}, \text{symmetric} \\
 & \quad \rho_{ij} \in \mathbb{C}^{4 \times 4}, \text{Hermitian} \forall i, j
 \end{aligned}$$



# What does the SDP give us?

- $M^*$  - optimal solution to SDP
  - Cholesky decomposition-  $\{v_{\sigma j}\}$  such that  $M_{\sigma j, \eta l} = v_{\sigma j}^T v_{\eta l}$ .
  - Statistics corresponding to a non-physical density matrix
  - However, objective of M **upper bounds** optimal density matrix
- $\{v_{\sigma j}\}$  represents unphysical statistics, can we **round to** a physical state with these vectors? Maybe a Bloch vector?
- Rounding to Bloch vectors would **guarantee** “physical-ness”



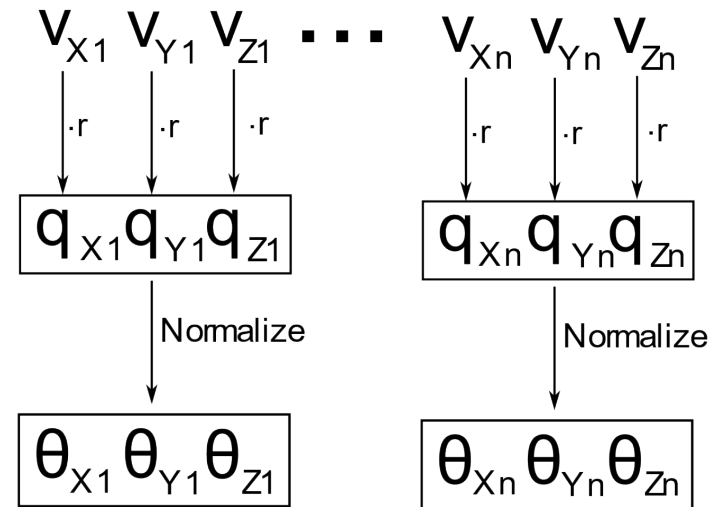
# “Hyperplane” Rounding

- Let  $r \sim \mathcal{N}(0, \mathbb{I})$ . If we could take  $\theta_{\sigma j} = v_{\sigma j} \cdot r$  :
  - $\mathbb{E}[(v_{\sigma j} \cdot r)(v_{\eta l} \cdot r)] = \mathbb{E}[v_{\sigma j}^T r r^T v_{\eta l}] = v_{\sigma j}^T v_{\eta l}$
- We would get approx. factor 1! But this is unphysical, need to normalize:

## Procedure

- Take  $q_{\sigma j} = v_{\sigma j} \cdot r \ \forall \sigma, j$
- Normalize:  $\theta_{\sigma j} = \frac{v_{\sigma j} \cdot r}{\sqrt{\sum_{\eta} (v_{\eta j} \cdot r)^2}}$
- Output:

$$\rho = \prod_i \left( \frac{\mathbb{I} + \sum_{\sigma} \theta_{\sigma} \sigma_i}{2} \right)$$



$$\theta_{X1} = \frac{q_{X1}}{\sqrt{q_{X1}^2 + q_{Y1}^2 + q_{Z1}^2}}$$



# Analysis



- Evaluate the expected approx. factor edge by edge:

$$\mathbb{E} \left[ \text{Tr} \left( \sum_e h_e \rho \right) \right] = \sum_e \mathbb{E} [\text{Tr}(h_e \rho)]$$

- If  $\frac{\mathbb{E}[\text{Tr}(h_e \rho)]}{\text{Tr}[M C^e]} \geq \alpha$  for all  $e$ , then  $\sum_e \mathbb{E}[\text{Tr}(h_e \rho)] \geq \alpha \sum_e \text{Tr}[M C^e]$
- Reduce edge to **standard form** which depends on the rank of  $h_e$
- Given this standard form, use **Hermite polynomials** to calculate the expectation:
  - $f(z), g(z)$  functions of normal r.v.'s  $\mathcal{N}(0, \mathbb{I})$
  - $\Rightarrow \mathbb{E}[f(z)g(z)] = \sum_{\mu} \hat{f}_{\mu} \hat{g}_{\mu}$
  - $\theta_{\sigma i}$  function of  $r \sim \mathcal{N}(0, \mathbb{I})$



# Future Work



- **Optimizing proof methods?** proven approximation factors are strict lower bounds on performance of alg., we suspect it is possible to prove actual performance with better techniques
- **1-Local vs. 2-Local?** We suspect alg. has same performance when 1-local terms are included
- **Genuinely entangled Ansatz?**
- **Approximation alg. making crucial use of quantum circuit?**

