

Class 3: PWF 2021 - QML 12/01/2021

$$C(\vec{\phi}) = \text{Tr} [U(\vec{\phi}) \rho_0 U^\dagger(\vec{\phi})]$$



O is usually of 2 forms

A) $O = \sum_i C_i \sigma_i$ $\sigma_i \in \{X, Y, Z, \text{ and } \text{set of Pauli strings}\}$

the summation has a polynomial number of terms

$$\underline{Z} \otimes \underline{Z} \otimes \underline{Y} \cdots \otimes \underline{Y} = Z_1 Z_2$$

$$Y_1 \otimes Z_2 \otimes Z_3 \otimes Z_4$$

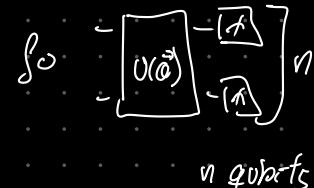
B) $O = \sum_i d_i |z_i\rangle \langle z_i|$ $|z_i\rangle \in \{|0\rangle, |1\rangle\}^{\otimes n}$

the summation has a polynomial number of terms

$$|0\cdots 0\rangle, |10\cdots 0\rangle, \dots |1\cdots 1\rangle$$

How do we estimate the cost through measurements that are usually projective measurements in the computational basis.

$$\begin{aligned} B) \Rightarrow C(\vec{\phi}) &= \text{Tr} [U(\vec{\phi}) \rho_0 U^\dagger(\vec{\phi})] \sum_i d_i |z_i\rangle \langle z_i| \\ &= \sum_i d_i \text{Tr} [U(\vec{\phi}) \rho_0 U^\dagger(\vec{\phi}) |z_i\rangle \langle z_i|] \\ &= \sum_i d_i \langle z_i | U(\vec{\phi}) \rho_0 U^\dagger(\vec{\phi}) | z_i \rangle \\ &= \sum_i d_i p(z_i) \approx \sum_i d_i \frac{N_{z_i}}{N} \end{aligned}$$



$$\sum_i d_i |z_i\rangle \langle z_i| \Rightarrow |0\cdots 0\rangle \langle 0\cdots 0|$$

$$C(\vec{\phi}) = p(0\cdots 0)$$



SWAP circuit

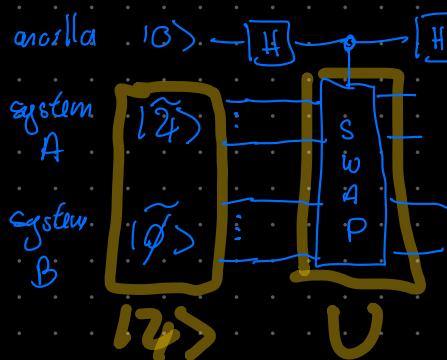


SWAP



$$\text{SWAP } |2\rangle \otimes |0\rangle = |0\rangle \otimes |2\rangle$$

Hadamard-Swap test



ancilla probability ancilla being in zero translating results

$$|2\rangle = |\tilde{2}\rangle \otimes |\tilde{0}\rangle$$

$$U = \text{SWAP}$$

$$|\tilde{2}_f\rangle = \frac{1}{2} \left(|0_A\rangle \otimes (|\tilde{2}\rangle|\tilde{0}\rangle + |\tilde{0}\rangle|\tilde{2}\rangle) + |1_A\rangle \otimes (|\tilde{2}\rangle|\tilde{0}\rangle - |\tilde{0}\rangle|\tilde{2}\rangle) \right)$$

$$\langle 2_f | U | 2 \rangle$$

$$P_A(0) = \frac{1}{2} [1 + \text{Re} \{ \langle \tilde{2} | \langle \tilde{0} | \text{SWAP} | \tilde{2} \rangle | \tilde{0} \rangle \}]$$

$$= \frac{1}{2} [1 + \text{Re} \{ \langle \tilde{2} | \langle \tilde{0} | \tilde{0} \rangle | \tilde{2} \rangle | \tilde{0} \rangle \}]$$

$$= \frac{1}{2} [1 + \text{Re} \{ \langle \tilde{2} | \tilde{0} \rangle \langle \tilde{0} | \tilde{2} \rangle \}]$$

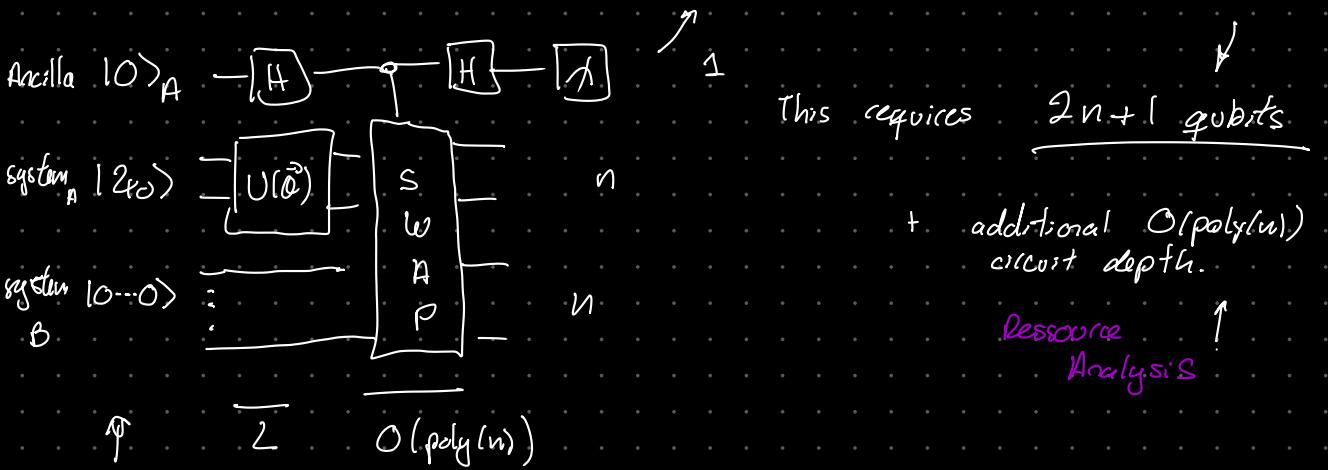
$$= \frac{1}{2} [1 + \text{Re} \{ | \langle \tilde{2} | \tilde{0} \rangle |^2 \}]$$

$$= \frac{1}{2} [1 + | \langle \tilde{2} | \tilde{0} \rangle |^2]$$

$$(A \otimes B) \cdot (C \otimes D) \\ = AC \otimes BD$$

Our cost function $C(\vec{\phi}) = \rho(O \dots O) = \langle 0 \dots 0 | U(\vec{\phi}) \mathcal{S}_0 U^\dagger(\vec{\phi}) | 0 \dots 0 \rangle$

$$\mathcal{S}_0 = |2_0 \times 2_0| = \underline{| \langle 0 \dots 0 | U(\vec{\phi}) | 2_0 \rangle |^2} \quad \textcircled{F}$$



This requires $2n+1$ qubits

+ additional $O(\text{poly}(n))$ circuit depth.

Resource Analysis

$\langle X_{\text{target}} \rangle$

$$C(\vec{\phi}) = 1 - \frac{|\langle X_{\text{target}} | 2_\phi \rangle|^2}{|\langle 2_\phi | 2_\phi \rangle|}$$

if $|2_\phi\rangle = e^{i\phi}|X_{\text{target}}\rangle$
then $C(\vec{\phi}) = 0$

A) $C = \sum_i C_i G_i$ $G_i \in \{X, Y, Z, 1\}^{\otimes n}$

$$\begin{aligned} C(\vec{\phi}) &= \sum_i C_i \text{Tr} [U(\vec{\phi}) | 2_\phi \rangle \langle 2_\phi | U^\dagger(\vec{\phi}) G_i] \\ &= \sum_i C_i \langle 2_\phi | U^\dagger(\vec{\phi}) G_i U(\vec{\phi}) | 2_\phi \rangle \\ &= \sum_i C_i \langle G_i |_{U(2_\phi)} \end{aligned}$$

we know that if $G_i = Z^{(k)}$

$$|2_\phi\rangle \xrightarrow{U(\vec{\phi})} \underbrace{\begin{matrix} \text{---} \\ | \end{matrix}}_{k\text{-th qubit}} \quad Z = \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix}$$

$$\langle Z^{(k)} \rangle = P_k(0) - P_k(1)$$

what happens if $G_i = X^{(k)}$

$$(2\omega) \quad \overline{\left[\begin{array}{c} \vdots \\ \text{U}(\vec{\phi}) \\ \vdots \end{array} \right]} - [\vec{x}]_x = \overline{\left[\begin{array}{c} \vdots \\ \text{U}(\vec{\phi}) \\ \vdots \end{array} \right]} - \overline{[\vec{H}]} - [\vec{x}]_z$$

$$\begin{aligned} P_k(\phi) &= \text{Tr} \left[\left(H^{(k)} \otimes \mathbb{1}_{\bar{k}} \right) \underbrace{U(\vec{\phi}) (2\omega \times 2\omega) U^f(\vec{\phi})}_{M_{\phi}^{(k)}} \left(H^{(k)} \otimes \mathbb{1}_{\bar{k}} \right) \left(10 \times d_k \otimes \mathbb{1}_{\bar{k}} \right) \right] \\ &= \text{Tr} \left[U(\vec{\phi}) (2\omega \times 2\omega) U^f(\vec{\phi}) \underbrace{\left(H^{(k)} \otimes \mathbb{1}_{\bar{k}} \right) \left(10 \times d_k \otimes \mathbb{1}_{\bar{k}} \right) \left(H^{(k)} \otimes \mathbb{1}_{\bar{k}} \right)}_{H^{(k)} 10 \times d_k H^{(k)} \otimes \mathbb{1}_{\bar{k}}} \right] \\ &\quad \underbrace{1 + X + I_k \otimes \mathbb{1}_{\bar{k}} \in M_{+}^{(k)}} \\ &= \text{Tr} \left[U(\vec{\phi}) (2\omega \times 2\omega) U^f(\vec{\phi}) M_{+}^{(k)} \right] \end{aligned}$$

$$\langle x \rangle = P_k(\phi) - P_k(I)$$

Now let's compute the expectation value of 2-body Pauli operators

$$\langle Z_1 Z_2 \rangle$$

$$\begin{aligned} Z_1 &= 10 \times d_1 - \| \vec{x} \|_1, & Z_1 Z_2 &= \left(10 \times d_1 - \| \vec{x} \|_1 \right) \otimes \left(10 \times d_2 - \| \vec{x} \|_2 \right) \\ Z_2 &= 10 \times d_2 - \| \vec{x} \|_2 & &= 100 \times d_1 d_2 - 10 \| \vec{x} \|_1 \| \vec{x} \|_2 - (10 \times d_1 \\ & & & \quad + \| \vec{x} \|_1 \| \vec{x} \|_2) \end{aligned}$$

$$\langle Z_1 Z_2 \rangle = P(00) + P(11) - P(01) - P(10) \quad \leftarrow$$



$$\langle Z_1 Z_2 \rangle \approx \frac{N_{00} + N_{11} - N_{01} - N_{10}}{N}$$



What happens $\langle Z_1 X_2 \rangle$

$$\begin{array}{c} \text{Top} \quad \downarrow \quad \downarrow \\ \langle Z_1 X_2 \rangle \approx \frac{N_{00} + N_{11} - N_{01} - N_{10}}{N} \end{array}$$

$|Z_0\rangle = \sum |U(\vec{O})\rangle$

Ressource analogsis

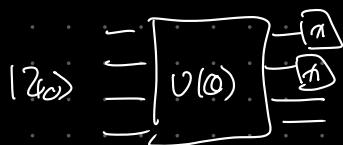
$$O = \sum_i C_i O_i$$

poly(n)-terms

estimating O_i requires a circuit
with (maybe) some charge ob-
asis, and local measurements
 N shots

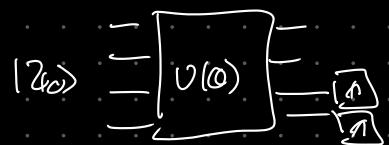
$\text{poly}(n) \times N$ measurement shots

Example I have a 4 qubit state $\langle Z_1 Z_2 \rangle \quad \langle Z_3 Z_4 \rangle$



$\langle Z_1 Z_2 \rangle$

N shots



$\langle Z_3 Z_4 \rangle$

N shots

+



$$p(0000) \quad p(1000) \quad p(0100)$$

$$\dots$$

$$p(1111)$$

$$2^4 = 16 \text{ outcomes}$$

$$p(x_i, y_j)$$

↓

$$p(x_i) = \sum_{y_j} p(x_i, y_j)$$

$$p(z_1 z_2) = p(0000) + p(0001) + p(0010) + p(0011)$$

This allows to compute both $\langle z_1 z_2 \rangle$ $\langle z_3 z_4 \rangle$ with N measurements?

$$z_1 z_2 = z_1 \otimes z_2 \otimes \mathbb{I}_3 \mathbb{I}_4 - \langle z_1 z_2, z_3 z_4 \rangle = 0$$

$$z_3 z_4 = \mathbb{I}_1 \otimes \mathbb{I}_2 \otimes z_3 \otimes z_4$$

$$x_2 x_3$$

$$\langle \underline{x_2 x_3}, \underline{z_1 z_2} \rangle \neq 0$$

$$\langle x_2 x_3, z_3 z_4 \rangle \neq 0$$

$$x_3 x_4$$

$$\langle z_1 z_2, x_3 x_4 \rangle = 0$$

+1, -1

$$\frac{1}{\sqrt{N}} \langle z_1 z_2 \rangle = \sum_{y_j} \overrightarrow{q^{(y_j)}_M} \overleftarrow{p(y_j)} \quad \begin{matrix} \downarrow \\ n-\text{length bitstrings} \end{matrix}$$

$$= \sum_{y_j} \overrightarrow{q^{(y_j)}_M} p(z_1 | z_i=i \text{ & } z_j=j)$$

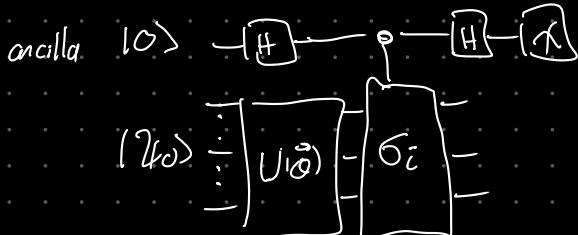
Error analysis:

$$\Delta(z_1 z_2) = \epsilon \Delta p$$

$$\Delta f(x_1 \dots x_n)$$

$$= \sum_i \left| \frac{\partial f}{\partial x_i} \right| \Delta x_i$$

Note: we can also compute $\langle G_i \rangle$ via the Hadamard test



$$\forall G_i \in \{X, Y, Z, H\}$$

$$G_i \cdot G_i^\dagger = G_i G_i = \mathbb{I}$$

$$z_1 z_2 - \bar{z}_1 \bar{z}_2$$

$$= z_1 z_2 \otimes \bar{z}_1 \bar{z}_2$$

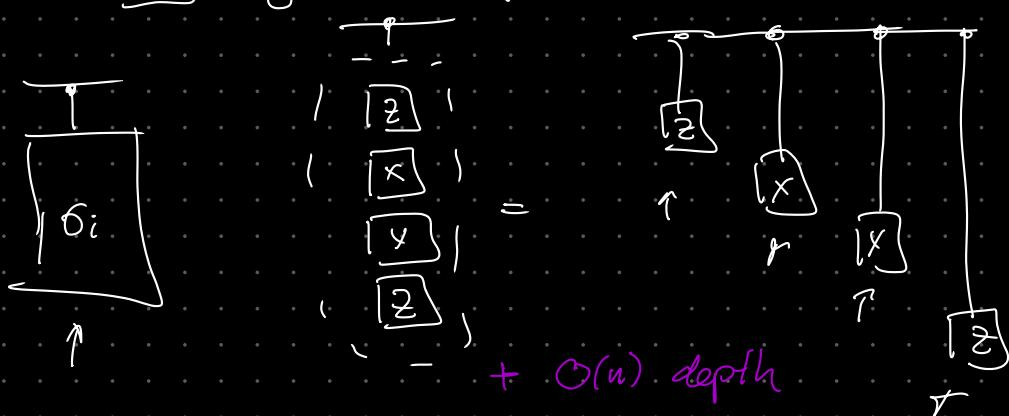
$$= \mathbb{I} \otimes \mathbb{I}$$

$$z_1 z_2$$

Assume G_i acts non-trivial on all n -qubits



if we use the hadamard test



$$(10 \times 0 \otimes \mathbb{I} + 11 \times \mathbb{I} \otimes X_2) \cdot (10 \times 0 \otimes \mathbb{I} + 11 \times \mathbb{I} \otimes Z_1)$$

$$(10 \times 0 \otimes H + 11 \times \mathbb{I} \otimes X_2 Z_1)$$

k -body Pauli string
 $O(n)$

QAOA : Quantum Approximation Optimization Algorithm

Farhi et al arxiv 1411.4028

Solving combinatorial problems in Quantum computers.

Used logistics, supply chains, developing networks
sending taxis & fixing rate, optimal delivering routes
water & electricity supply

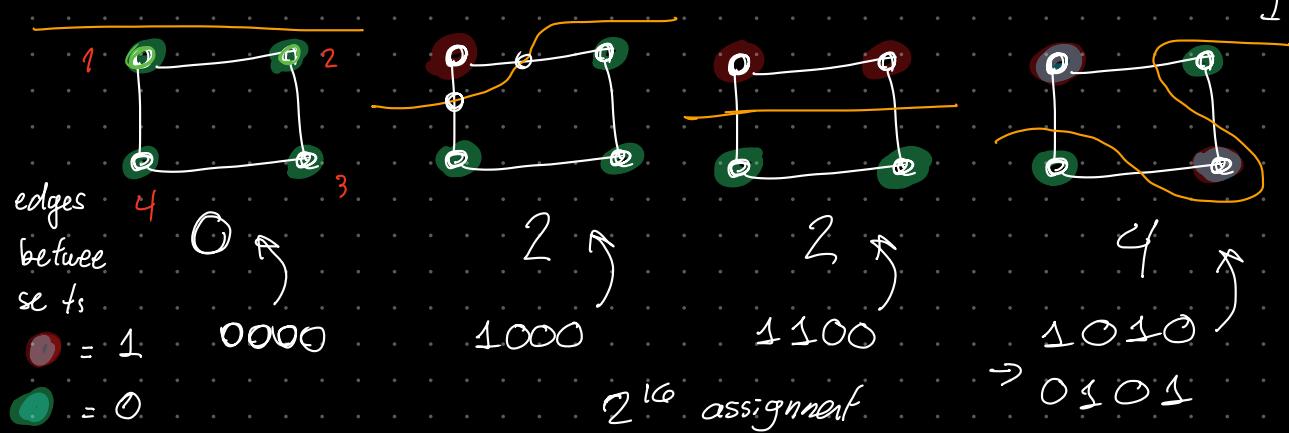
General : Hard to solve

formally speaking a Combinatorial optimization is defined as a task of determining optimal object out of a finite set of objects.

We will focus: we find an optimal bitstring, optimal sequence of 0's and 1's.

Graphs \rightarrow Max-cut problem

Partitioning nodes on a graph into 2 sets, such that the number of edges between the sets is maximum.



As the # of nodes increases, the number of possible assignments increases exponentially.

i) we need to define a cost function

Maxcut problems are defined of a graph

$$G = (V, E)$$

nodes ↴ ↵ edges

Object function for the classical problem is defined on bitstrings of length n .

$$S(z) = \sum_{\alpha=1}^E S_\alpha(z) \quad z \in \{0, 1\}^n$$

here $S_\alpha(z) = 1$ if z satisfies a clause

$S_\alpha(z) = 0$ otherwise

we want to promote the classical object function a quantum cost function

$$\boxed{H_P = \sum_{\langle jk \rangle} H_{jk}} \quad \begin{matrix} \nearrow & \searrow \\ \text{edges on the graph} & \end{matrix}$$

where

$$H_{jk} = \frac{1}{2} (1 - Z_j Z_k)$$

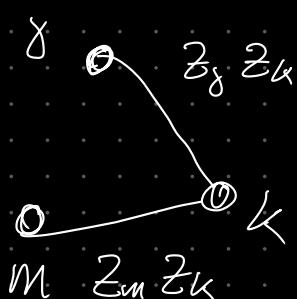
$$\begin{matrix} 1 & -1 \\ \downarrow & \uparrow \\ 0 & 1 \end{matrix}$$

$$\left. \begin{matrix} 1 \\ 1 \end{matrix} \right\} \begin{matrix} k\text{-th qubit is in } 0 \\ j\text{-th qubit is in } 0 \end{matrix} \quad \left. \begin{matrix} 1 \\ 1 \end{matrix} \right\} \begin{matrix} 1 \\ 1 \end{matrix}$$

$$\left[\langle z | H_{jk} | z \rangle = 0 \right]$$

$$\left. \begin{matrix} 1 \\ 1 \end{matrix} \right\} \begin{matrix} \text{a-th qubit is in } 0 \\ j\text{-th qubit is in } 1 \end{matrix} \quad \left. \begin{matrix} 1 \\ 0 \end{matrix} \right\} \begin{matrix} 1 \\ 0 \end{matrix}$$

$$\left[\langle z | H_{jk} | z \rangle = 1 \right]$$



The QAOA cost function

$$C(\vec{\phi}) = \text{Tr} \left[U(\vec{\phi}) | \Psi_0 \rangle \langle \Psi_0 | U^\dagger(\vec{\phi}) H_P \right]$$

2) Ansatz for $U(\vec{\phi})$ loosely base on the adiabatic theorem

$$\begin{array}{ccc} H_M & \swarrow & H_P \\ |GS_M\rangle & \xrightarrow{\quad} & |GS_P\rangle \\ \downarrow & & \uparrow \\ e^{-iH(t)\epsilon} & \text{hard to implement} & s(0) = 0 \\ & & s(t_f) = 1 \end{array}$$

$$H(t) = (1-s(t))H_M + s(t)H_P$$

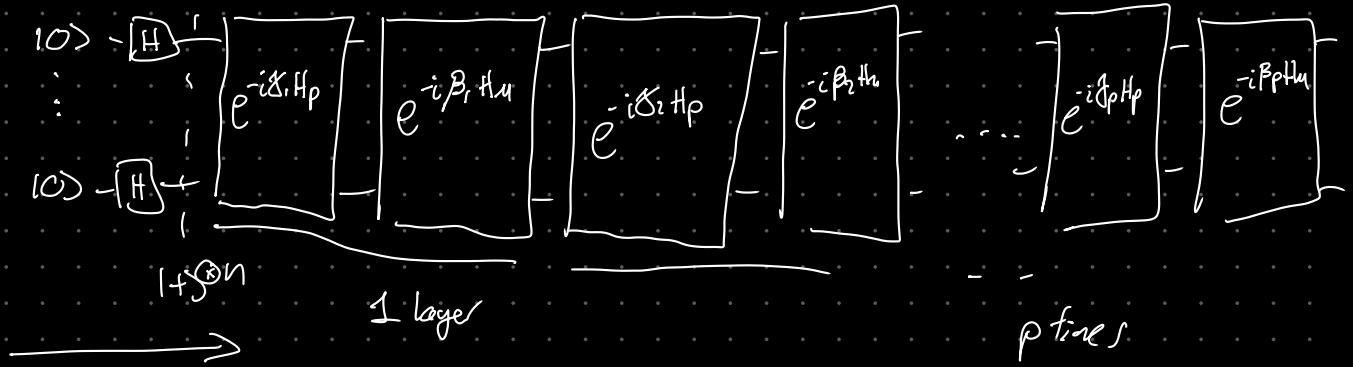
i) Discretize t into P , Δt intervals.

$$\begin{aligned} U(t_f, \phi) &= U(t_f, t_f - \Delta t) \dots U(0, \phi) \\ &= \prod_{j=1}^P U(j\Delta t, (j-1)\Delta t) \\ &\approx e^{-i \sum_j H_j \Delta t} \end{aligned}$$

$$e^{i(A+B)x} \approx e^{Ax} e^{iBx} + O(x^2)$$

$$U(t_f, \phi) = \prod_{j=1}^P e^{-i(1-s_j \Delta t)} \underbrace{H_M \Delta t}_{\uparrow} e^{-i \sum_j (g_j \phi) \frac{H_P \Delta t}{\uparrow}}$$

$$\text{Or } \begin{aligned} \text{"mixed"} &= H_M = \sum_i X_i & |GS_M\rangle &= |+\rangle^{\otimes n} \\ \text{"driven"} &= \uparrow \end{aligned}$$



$$U(\vec{\beta}, \vec{\gamma}) = U(H_m, \beta_p) U(H_p, \beta_p) \cdots U(H_m, \beta_1) U(H_p, \beta_1)$$

$\uparrow \quad \uparrow \quad \uparrow \quad \uparrow \quad \uparrow \quad \uparrow$

$$U(\vec{\beta}, \vec{\gamma})(\Psi_0) =$$

$$\underbrace{A \cdot B \cdot (\chi)}$$

③ How do we compute the cost ?

$$\begin{aligned}
 C(\theta) &= \sum_{j,k} \langle H_{jk} \rangle \\
 &= \sum_{jk} \langle H_{jk} \rangle \quad \downarrow \\
 &= \sum_{jk} \frac{1}{2} (1 - \langle Z_j Z_k \rangle)
 \end{aligned}$$

$H_{jk} = \frac{1}{2} (1 - Z_j Z_k)$

$$\langle Z_j Z_k, Z_m Z_p \rangle = 0$$

$$\# j, k, m, p$$

$$\langle X_i, X_j \rangle = 0 \quad \forall i, j$$

$$e^{A+B} = e^A e^B$$

$$\text{if } \{A, B\} = 0$$

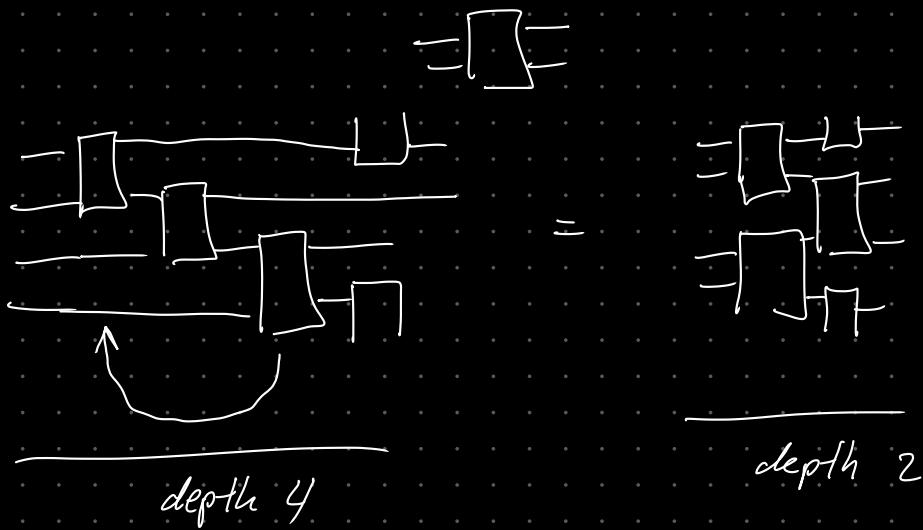
2.1

$$\begin{aligned}
 e^{-i H_m \beta} &= e^{-i \beta} \sum_i X_i \\
 &= \prod_{i=1}^n \frac{e^{-i \beta} X_i}{R_X(\beta)}
 \end{aligned}$$

$\cancel{-[R_X]} \quad \cancel{-[R_X]} \quad \cancel{-[R_X]}$
 same param. β

$$\begin{aligned}
 e^{-i\gamma H_p} &= e^{-i\gamma \sum_{j \in S} \frac{1}{2} (1 - z_j z_{ja})} & \sum_{j \in S} \neq \sum_{j \in C} \\
 &\equiv e^{-i\gamma/2} \underbrace{1}_{\perp} + i\gamma \sum_{j \in S} \overline{z_j z_{ja}} \underbrace{\frac{1}{2}}_{\perp} \\
 &= e^{-i\gamma/2} \underbrace{\pi}_{\langle j \in C \rangle} e^{i\gamma/2} \frac{\overline{z_j z_{ja}}}{\langle j \in C \rangle}
 \end{aligned}$$

$$\begin{array}{c}
 \text{j-th qubit} \\
 \text{k-th qubit}
 \end{array}
 \quad
 \begin{array}{c}
 \text{Circuit Diagram} \\
 \text{with } R_\theta(\alpha) \text{ gate}
 \end{array}
 = e^{-i\alpha z_j z_{ja}}$$



The generalization of the ansatz of $\psi(\alpha)$.

Hadfield et al. Algorithms (2022) 34 (2019)
arXiv 1709.0348

Given H_u, H_p
(find $|GS\rangle$)

QAOA = Quantum Alternating Operator Ansatz

Quantum Autoencoder

$$|\Psi_i\rangle$$

n-qubit quantum

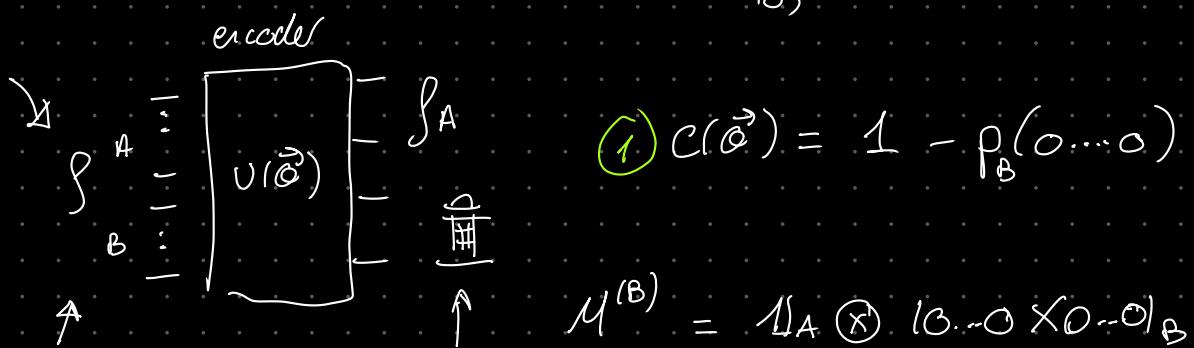
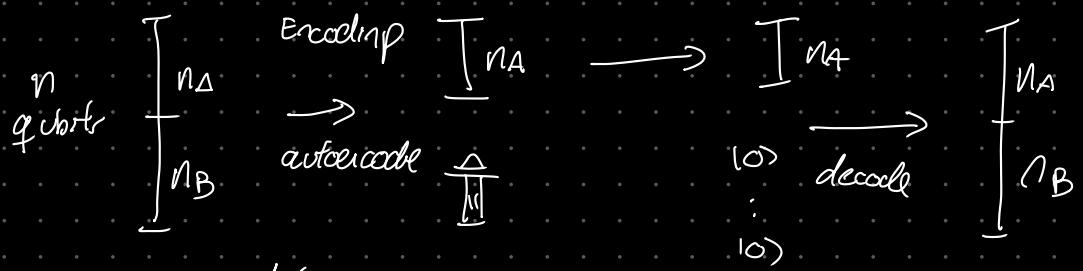
$$\rho = \sum_i p_i |\Psi_i\rangle \langle \Psi_i|$$

Romeo et al

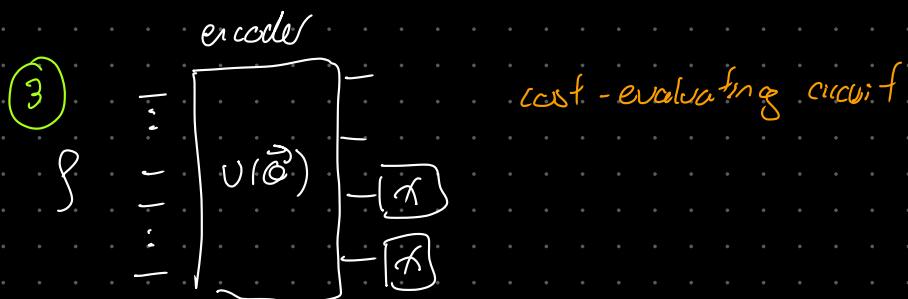
Quantum Sci Tech

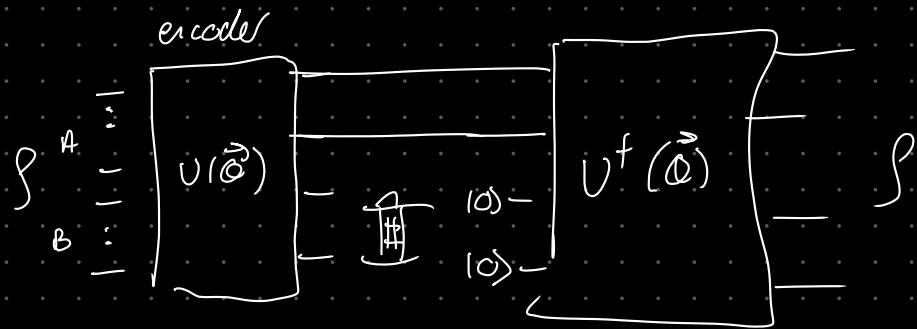
2045001, (2017)

arXiv/1612.02806

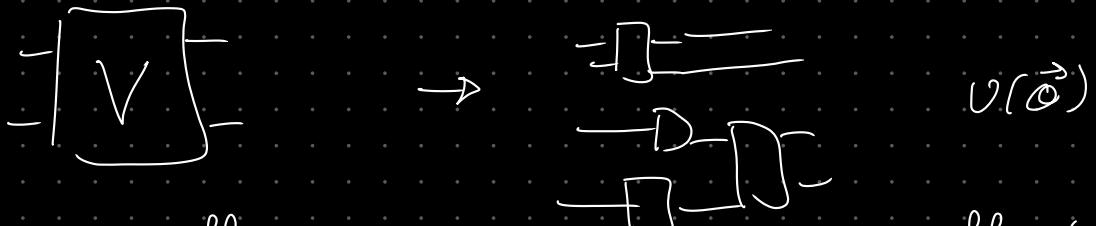


$$C(\vec{O}) = 1 - p_B(0\ldots 0) = \text{Tr} \left[U(\vec{O}) \rho U^\dagger(\vec{O}) (\mathbb{1}_A \otimes |0\ldots 0\rangle \langle 0\ldots 0|_B) \right]$$





Compilation Algorithm



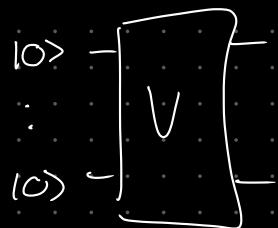
- has no efficient quantum circuit description

Task we learn an efficient quantum circuit representation of a unitary V

Goal

- unknown

$$e^{-iHt}$$



$$V|O\rangle$$



$$\frac{1}{\sqrt{2}}$$

$$\frac{1}{\sqrt{2}}$$



$$U(O)|O\rangle$$



we want $U(O)|O\rangle$ to match $V|O\rangle$

①
③

$$C(O) = 1 - |\langle O | U^f(O) | V | O \rangle|^2$$

$$C(\vec{\phi}) = 1 - \frac{|\langle \vec{\phi} | U^\dagger(\vec{\phi}) V |\vec{\phi} \rangle|^2}{\langle \vec{\phi} | \vec{\phi} \rangle}$$

How do we evaluate $C(\vec{\phi})$: 1 SWAP TEST

$$\begin{aligned} |0\rangle & - \left| \begin{array}{c} \text{V} \\ - \end{array} \right\rangle \xrightarrow{U^\dagger(\vec{\phi})} \left| \begin{array}{c} \text{X} \\ - \end{array} \right\rangle \\ |0\rangle & \leftarrow \left| \begin{array}{c} \text{V} \\ - \end{array} \right\rangle \xrightarrow{U(\vec{\phi})} \left| \begin{array}{c} \text{X} \\ \text{X} \end{array} \right\rangle \end{aligned}$$

$$C(\vec{\phi}) = 1 - P(\vec{\phi})$$

- ② For the autoencoder & the compilation test:
what ansatz should we use for $U(\vec{\phi})$?

Ansatz design
Problem