

A QUANTUM ALGORITHM FOR THE SUB-GRAPH ISOMORPHISM PROBLEM

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Sub-graph isomorphism (SGI)

We consider an (*finite*) *undirected graph* $\mathcal{A} = (V_{\mathcal{A}}, E_{\mathcal{A}})$, where

- $V_{\mathcal{A}}$ is the finite set of vertices,
- $E_{\mathcal{A}}$ is the set of elements of the form $\{i, j\}$, representing the edge connecting vertex $i \in V_{\mathcal{A}}$ to $j \in V_{\mathcal{A}}$.

To each graph \mathcal{A} , with $n = |V_{\mathcal{A}}|$, we associate the corresponding $n \times n$ *adjacency matrix* A whose entries are

$$A_{i,j} = \begin{cases} 1, & \{i, j\} \in E_{\mathcal{A}} \\ 0, & \text{otherwise,} \end{cases}$$

A graph \mathcal{B} is said to be an *induced subgraph* of a graph \mathcal{A} if $V_{\mathcal{B}} \subset V_{\mathcal{A}}$ and for all $i, j \in V_{\mathcal{B}}$, $\{i, j\} \in E_{\mathcal{A}} \iff \{i, j\} \in E_{\mathcal{B}}$.

Let $n = |V_{\mathcal{A}}|$, $m = |V_{\mathcal{B}}|$, then the sub-graph isomorphisms are the $(n \times n)$ permutations P that make the following cost function vanish

$$\ell_{A,B}(P) = \|SPAP^{\top}S^{\top} - B\|_F^2,$$

where $S = (\mathbb{I}_m \mid \mathbb{O}_{m,n-m})$. The space of m -permutation of n has size $\frac{n!}{(n-m)!}$, which becomes $n!$ when $n = m$.

The hat representation $\widehat{\cdot}$

Given graph \mathcal{A} and the corresponding adjacency matrix $A \in \mathcal{S}_N(\mathbb{Z}_2)$, with $N = 2^k = |V_{\mathcal{A}}|$ for some $k > 1$, we define the *unitary operator representation* denoted $\widehat{A} \in \mathbf{U}(2N^2)$, as

$$\widehat{A} := H^{\otimes(2k+1)} \left(\mathbb{I}_{N^2} \oplus \sum_{i,j} (-1)^{A_{i,j}} |i, j\rangle \langle i, j| \right) H^{\otimes(2k+1)}, \quad (1)$$

where H is the 2×2 matrix corresponding to the Hadamard operator on a single qubit.

Given the $N \times N$ permutation P we define the representation

$$\check{P} := \mathbb{I}_2 \otimes (H^{\otimes 2k} \cdot P^{\otimes 2} \cdot H^{\otimes 2k}). \quad (2)$$

Properties of $\widehat{\cdot}$ and $\check{\cdot}$

For all $A, B \in \mathcal{S}_N(\mathbb{Z}_2)$,

- (i) $\widehat{A +_2 B} = \widehat{A} \cdot \widehat{B} = \widehat{B} \cdot \widehat{A}$, $\widehat{\mathbb{O}_N} = \mathbb{I}_{2N^2}$ (group-homomorphism)
- (ii) $A \neq B \implies \widehat{A} \neq e^{i\phi} \widehat{B}$, $\forall \phi \in \mathbb{R}$ ('physical injectivity')
- (iii) $\widehat{PAP^{\top}} = \check{P} \cdot \widehat{A} \cdot \check{P}^{\top}$

Also $\widehat{\cdot}$ requires $\log_2(2N^2) = 2k + 1$ qubits w.r.t. the number of vertices N .

The Ansatz

Given a vector of parameters $\theta \in \mathbb{R}^c$, we propose an Ansatz of the form

$$\tilde{P}_{\mathfrak{G}}(\theta) = e^{i\phi} \prod_{i=1}^c \exp\left(-i\frac{\theta_i}{2} P_i\right), \quad (3)$$

where P_i are permutation matrices acting on k qubits, such that $P_i^2 = \mathbb{I}_{2^k}$. Also, the structure of the P_i is determined by the hyperparameters \mathfrak{G} . Equation (3) can be expanded to reveal the tunable superposition of permutations structure

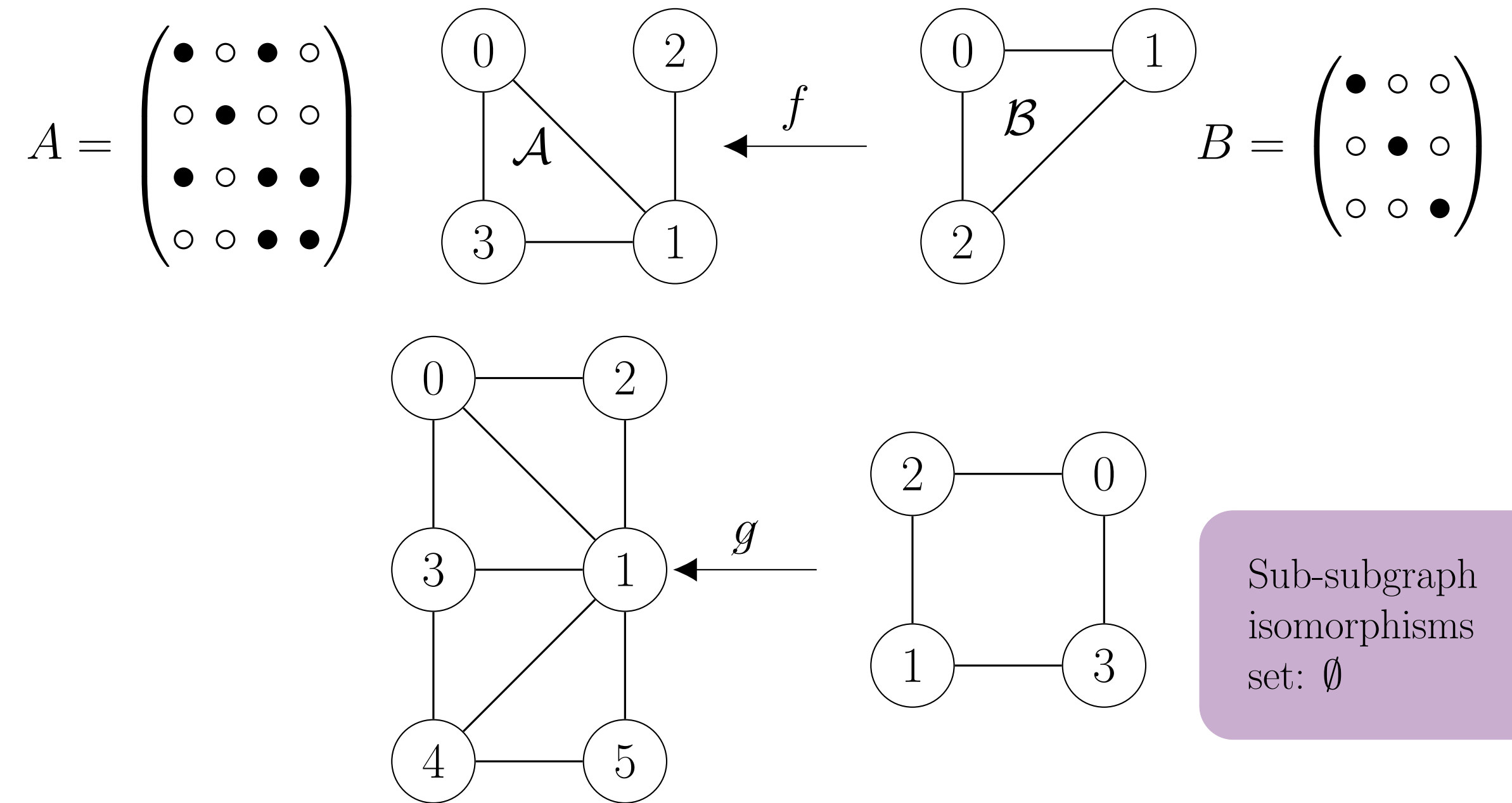
$$\tilde{P}_{\mathfrak{G}}(\theta) = \sum_{j=1}^{2^c} \alpha_j(\theta) Q_j,$$

where Q_j are permutation matrices (not necessarily unique) and the functions $\alpha_j : \mathbb{R}^c \rightarrow \mathbb{C}$ map the parameters to the complex coefficients of the Q_j . Moreover, we show that the sub-set of parameters $\{\theta \in \mathbb{R}^c | \theta_i \in \{0, \pi\}\}$ determines $\tilde{P}_{\mathfrak{G}}(\theta)$ to produce each Q_i individually.

Algorithm 1: The sub-graph isomorphism algorithm (simplified)

Data: Adjacency matrices A and B for graphs \mathcal{A} and \mathcal{B} , respectively. Permutation Ansatz hyper-parameters \mathfrak{G} .
// The matrices A and B have sizes $N \times N$ and $N_B \times N_B$, $N = 2^k$, $N_B = 2^{k'}$, $k' < k$
Result: R ; **// Set of partial permutation matrices**
 $R \leftarrow \emptyset$;
 $\mathcal{L}_Q^{\text{SGI}}(\theta) \leftarrow \text{CircuitFactory}(A, B, \mathfrak{G})$;
 $\theta \leftarrow \text{SampleUniform}([0, \pi]^c)$; **// Sample initial $\theta \in \mathbb{R}^c$ from $[0, \pi]^c$**
for $t_1 \leftarrow 1$ **to** maxsteps **do**
 $\theta \leftarrow \theta - \eta \tilde{\nabla}^{(t_1)}(-\mathcal{L}_Q^{\text{SGI}}(\theta))$; **// Update θ with SGD step**
 $\mathbf{p} \leftarrow (\Lambda(\theta_i/\pi))_i$; **// Obtain vector of distances of θ_i/π from closest even integer**
 for $t_2 \leftarrow 1$ **to** samples **do**
 $\mathbf{u} \leftarrow \text{SampleUniform}([0, 1]^c)$;
 $\mathbf{g} \leftarrow (\chi_{[0,\infty)}(p_i - u_i))_i$; **// Sample vector in $\{0, 1\}^c$ according to probabilities \mathbf{p}**
 $P \leftarrow \tilde{P}_{\mathfrak{G}}(\pi \cdot \mathbf{g})$; **// Obtain a classical permutation, note $\pi \cdot \mathbf{g} \in \{0, \pi\}^c$**
 $d \leftarrow \|SPAP^{\top}S^{\top} - B\|_F^2$; **// Check discrepancy classically**
 if $d = 0$ **then** $R \leftarrow R \cup \{SP\}$;
 end
 if $|R| \neq 0$ **then break**;
end

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Applications

Graph databases, biochemistry, computer vision, social network analysis, knowledge graph query and anti-money laundry.

The cost function

For simplicity consider the *graph isomorphism* (GI) case. We define the *classical disparity function* $\Psi_C : \mathcal{S}_N(\mathbb{Z}_2) \times \mathcal{S}_N(\mathbb{Z}_2) \rightarrow \mathbb{R}$ as

$$\Psi_C(A, B) := \frac{1}{N^2} \|A - B\|_F^2 = \frac{1}{N^2} \sum_{i,j} (A_{i,j} - B_{i,j})^2. \quad (4)$$

We define the corresponding *quantum disparity function* as

$$\Psi_Q^{\text{GI}}(A, B) := 1 - \left| \langle 0 |^{\otimes(2k+1)} \widehat{B} \cdot \widehat{A} | 0 \rangle^{\otimes(2k+1)} \right|, \quad (5)$$

also we prove that $\Psi_C(A, B) = \Psi_Q^{\text{GI}}(A, B)$.

For the SGI case we obtain the corresponding disparity function $\Psi_Q^{\text{SGI}}(A, B)$, not requiring $|V_{\mathcal{A}}| = |V_{\mathcal{B}}|$ also we prove again that $\Psi_C(SAS^{\top}, B) = \Psi_Q^{\text{SGI}}(A, B)$.

Now we introduce a permutation P to obtain

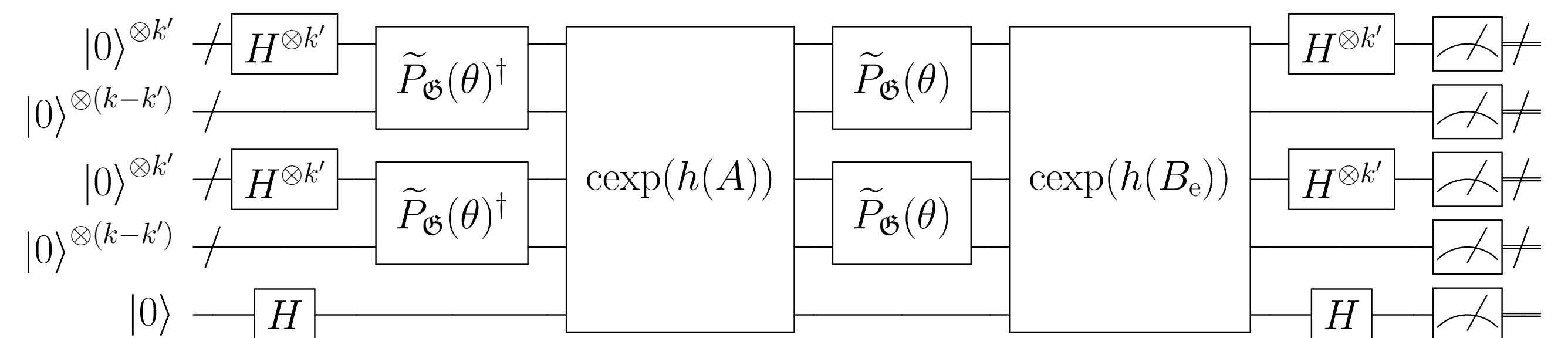
$$\Psi_Q^{\text{GI}}(PAP^{\top}, B) := 1 - \left| \langle 0 |^{\otimes(2k+1)} \widehat{B} \cdot \check{P} \cdot \widehat{A} | 0 \rangle^{\otimes(2k+1)} \right|. \quad (6)$$

Next step

We substitute the fixed permutation P with an Ansatz $\tilde{P}_{\mathfrak{G}}(\theta)$ producing a **tunable superposition of permutations**.

The quantum circuit

Consider graphs \mathcal{A} and \mathcal{B} having $|V_{\mathcal{A}}| = 2^k$ and $|V_{\mathcal{B}}| = 2^{k'}$ vertices, respectively ($k > k'$). Given the Ansatz $\tilde{P}_{\mathfrak{G}}(\theta)$, we obtain the QC for the SGI utility function $\mathcal{L}_Q^{\text{SGI}}(\theta)$.



Where

$$\text{cexp}(h(A)) = \mathbb{I}_{N^2} \oplus \sum_{i,j} (-1)^{A_{i,j}} |i, j\rangle \langle i, j|$$

is a *diagonal operator* depending on the adjacency matrix A .

Results

Below a practical example of SGI (columns 1-3) and two samples of the SGD evolution (fourth column).

