

1 Efficient simulation of commutators

H_1, H_2 are Hamiltonians of two n -qubit systems.

(a) Prove $e^{-[H_1, H_2]t} = \lim_{m \rightarrow \infty} (e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}})^m$

Proof.

$$\begin{aligned}
 A(m) &:= \left\| \left(e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}} \right)^m - e^{-[H_1, H_2]t} \right\| \\
 &= \left\| \left(e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}} \right)^m - \left(e^{\frac{-H_1 H_2 t + H_2 H_1 t}{m}} \right)^m \right\| \\
 &\leq m \left\| e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}} - e^{-H_1 H_2 t/m + H_2 H_1 t/m} \right\| \\
 &\cdot \max \left\{ \left\| e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}} \right\|, \left\| e^{-H_1 H_2 t/m + H_2 H_1 t/m} \right\| \right\}^{m-1} \\
 &\quad e^{\pm iH_1 \sqrt{t/m}}, e^{\pm iH_2 \sqrt{t/m}} \text{ are unitaries} \\
 &\Rightarrow \left\| e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}} \right\| \leq 1 \\
 &\left\| e^{-H_1 H_2 t/m + H_2 H_1 t/m} \right\| \leq e^{\| -H_1 H_2 t/m + H_2 H_1 t/m \|} = e^{\frac{t}{m} \| -H_1 H_2 + H_2 H_1 \|} \\
 &\quad e^{-iH_1 \sqrt{t/m}} e^{-iH_2 \sqrt{t/m}} e^{iH_1 \sqrt{t/m}} e^{iH_2 \sqrt{t/m}} - e^{-H_1 H_2 t/m + H_2 H_1 t/m} \\
 &= (I - iH_1 \sqrt{t/m} - \frac{1}{2} H_1^2 t/m + O(\|H_1\|^4 t^2/m^2)) (I - iH_2 \sqrt{t/m} - \frac{1}{2} H_2^2 t/m + O(\|H_2\|^4 t^2/m^2)) \\
 &\quad (I + iH_1 \sqrt{t/m} - \frac{1}{2} H_1^2 t/m + O(\|H_1\|^4 t^2/m^2)) (I + iH_2 \sqrt{t/m} - \frac{1}{2} H_2^2 t/m + O(\|H_2\|^4 t^2/m^2)) \\
 &\quad - (I - H_1 H_2 t/m + H_2 H_1 t/m + O(\|H_1 H_2 - H_2 H_1\|^2 t^2/m^2)) \\
 &= I - I - iH_1 \sqrt{t/m} - iH_2 \sqrt{t/m} + iH_1 \sqrt{t/m} + iH_2 \sqrt{t/m} - \frac{1}{2} H_1^2 t/m - \frac{1}{2} H_2^2 t/m \\
 &\quad - \frac{1}{2} H_1^2 t/m - \frac{1}{2} H_2^2 t/m - H_1 H_2 t/m + H_1^2 t/m + H_1 H_2 t/m + H_2 H_1 t/m + H_2^2 t/m \\
 &\quad - H_1 H_2 t/m + H_1 H_2 t/m - H_2 H_1 t/m + O(\max \{ \|H_1 H_2 - H_2 H_1\|^2, \|H_1\|^4, \|H_2\|^4 \} t^2/m^2) \\
 &= O(\max \{ \|H_1 H_2 - H_2 H_1\|^2, \|H_1\|^4, \|H_2\|^4 \} t^2/m^2)
 \end{aligned}$$

We have $A(m) \leq e^{\frac{t}{m} \|H_1 H_2 - H_2 H_1\|} O(\max \{ \|H_1 H_2 - H_2 H_1\|^2, \|H_1\|^4, \|H_2\|^4 \} t^2/m^2)$

Let $m \rightarrow \infty$, we are done. #

(b) H_1, H_2 can be efficiently simulated, then $i[H_1, H_2]$ is Hermitian and can be efficiently simulated.

Proof. $(i[H_1, H_2])^\dagger = (iH_1H_2 - iH_2H_1)^\dagger = -iH_1H_2 + iH_2H_1 = i[H_1, H_2]$

So $i[H_1, H_2]$ is Hermitian

We also have $A(m) = \left\| e^{-[H_1, H_2]t} \left(e^{-iH_1\sqrt{t/m}} e^{-iH_2\sqrt{t/m}} e^{iH_1\sqrt{t/m}} e^{iH_2\sqrt{t/m}} \right)^m \right\|$
 $\leq e^{\frac{t}{m} \|H_1H_2 - H_2H_1\|} O\left(\max\{\|H_1\|^4, \|H_2\|^4, \|H_1H_2 - H_2H_1\|^2\} \frac{t^2}{m}\right)$

To have $A(m) \leq \epsilon$, it suffices to let

$$m = O\left(\frac{1}{\epsilon} t^2 \max\{\|H_1\|^4, \|H_2\|^4, \|H_1\|^2 \|H_2\|^2\}\right) + O\left(t \|H_1H_2 - H_2H_1\| \frac{1}{\log 2}\right)$$

And each $e^{-iH_1\sqrt{t/m}} e^{-iH_2\sqrt{t/m}} e^{iH_1\sqrt{t/m}} e^{iH_2\sqrt{t/m}}$ can be implemented in $\text{poly}(n, t, \frac{1}{\epsilon})$

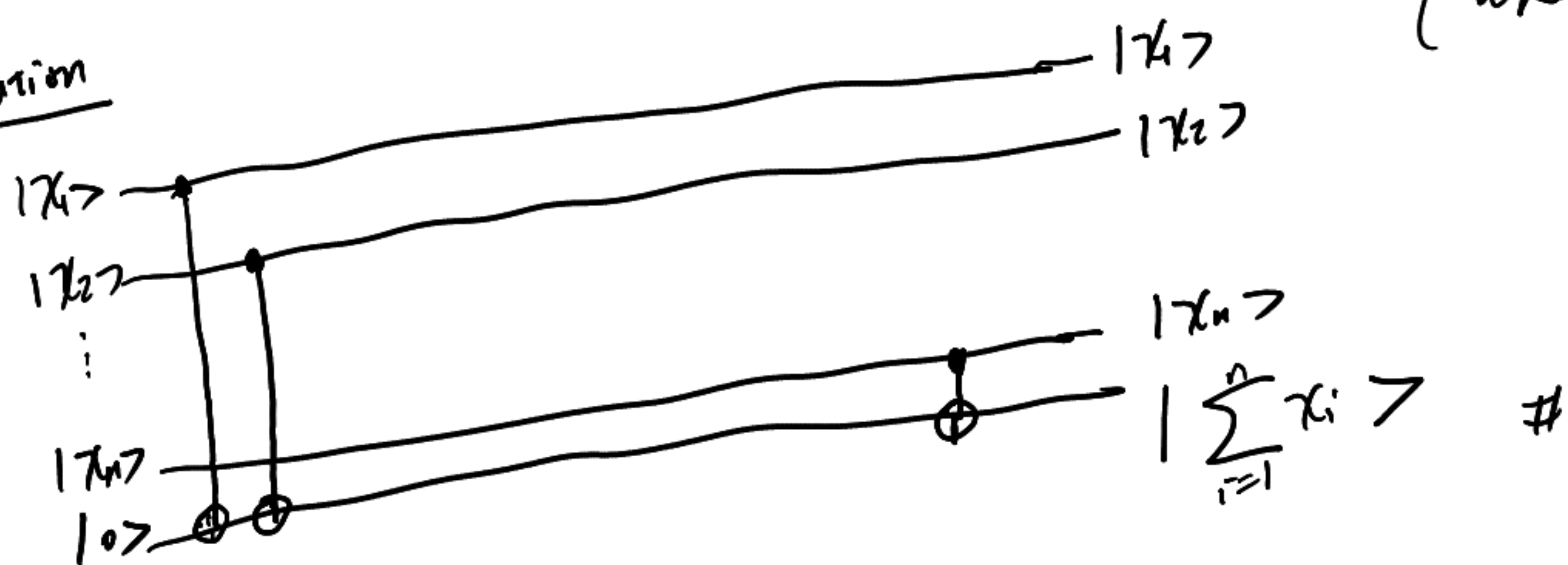
As $\|H_1\| = \text{poly}(n)$, $\|H_2\| = \text{poly}(n)$, $m = \text{poly}(n, t, \frac{1}{\epsilon})$

So total cost is $\text{poly}(n, t, \frac{1}{\epsilon})$ to achieve error $\leq \epsilon$. #

2 Simulation of Pauli Hamiltonians

(a) Design circuit for $|x\rangle|0\rangle \rightarrow |x\rangle|\sum_{i=1}^n x_i \bmod 2\rangle$, $\forall x \in \{0,1\}^n$
with $O(n)$ 1-qubit and 2-qubit gates

Solution



(b) $P = Z \otimes \dots \otimes Z$ what is $P|x\rangle$ where $|x\rangle$ is a basis?

Solution

$$P|x_1 \dots x_n\rangle = Z|x_1\rangle \otimes \dots \otimes Z|x_n\rangle = (-1)^{\sum_{i=1}^n x_i} |x_1 \dots x_n\rangle$$

where $x_i = 0$ or 1 , because $\begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$, $\begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$. #

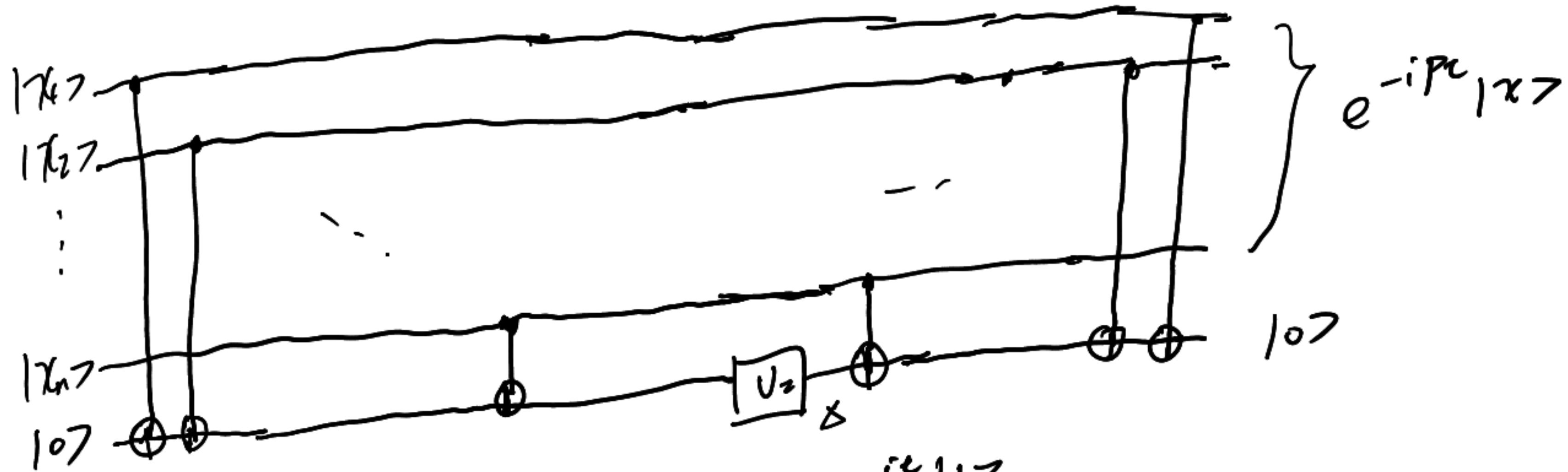
(c) $U = e^{-iPt}$ what is $U|x\rangle$ where $|x\rangle$ is a basis?

Solution

$$\begin{aligned} U|x_1 \dots x_n\rangle &= e^{-iPt} |x_1 \dots x_n\rangle = \sum_{k=0}^{\infty} \frac{(-i)^k t^k}{k!} P^k |x_1 \dots x_n\rangle \\ &= \sum_{k=0}^{\infty} \frac{(-i)^k t^k}{k!} (-1)^k \sum_{i=1}^n x_i |x_1 \dots x_n\rangle = \sum_{k=0}^{\infty} \frac{(-it(-1)^{\sum_{i=1}^n x_i})^k}{k!} |x_1 \dots x_n\rangle \\ &= e^{-it(-1)^{\sum_{i=1}^n x_i}} |x_1 \dots x_n\rangle, \text{ where } x_i = 0 \text{ or } 1. \# \end{aligned}$$

(d) Implement U with $O(n)$ 1-qubit gates and 2-qubit gates with $|0\rangle$ as auxiliary gate.

Solution



where $U_2|0\rangle = e^{-it}|0\rangle$, $U_2|1\rangle = e^{it}|1\rangle$.

Correctness:

At point Δ , we get $\begin{cases} e^{-it} |x\rangle |0\rangle & \text{if } \sum x_i = 0 \\ e^{it} |x\rangle |1\rangle & \text{if } \sum x_i = 1 \end{cases}$

Finally, we get $\begin{cases} e^{-it} |x\rangle |0 + \sum x_i\rangle = e^{-it} |x\rangle |0\rangle & \text{if } \sum x_i = 0 \\ e^{it} |x\rangle |1 + \sum x_i\rangle = e^{it} |x\rangle |0\rangle & \text{if } \sum x_i = 1 \end{cases}$

#

3 Estimating ground state energy

Given n -qubit $|\psi\rangle$ with $|\langle\psi|\psi_{\min}\rangle|^2 \geq 0.7$ $H = \sum_i a_i U_i$, U_i acts on 2-qubit systems. Output λ_{\min} with precision of $\lfloor 2 \log_2 n \rfloor$ bits with $\text{poly}(n)$ gates with probability $\geq \frac{2}{3}$.

Solution. wlog, $\text{Spec}(H) = \{\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_{2^n}\}$, $0 \leq \lambda_i < 1$.

$$H = \sum_{i=1}^{2^n} \lambda_i |\lambda_i\rangle \langle \lambda_i|$$

$$e^{-iHt} = \sum_{j=1}^{2^n} e^{-i\lambda_j t} |\lambda_j\rangle \langle \lambda_j|$$

$$= e^{-i\lambda_1 t} |\lambda_1\rangle \langle \lambda_1| + \sum_{j=2}^{2^n} e^{-i\lambda_j t} |\lambda_j\rangle \langle \lambda_j|$$

Since $|\langle\psi|\lambda_1\rangle|^2 \geq 0.7$

Expand $|\psi\rangle = b_1 |\lambda_1\rangle + \sum_{j=2} b_j |\lambda_j\rangle$ We have $\|b\|^2 \geq 0.7$, $b_j \in \mathbb{C}$, $j \geq 1$

$$e^{-iHt} |\psi\rangle = b_1 e^{-i\lambda_1 t} |\lambda_1\rangle + \sum_{j=2} c_j |\lambda_j\rangle, \quad c_j \in \mathbb{C}.$$

We know H is sparse, can be efficiently simulated

$$\text{Let } \|U - e^{-iHt}\| \leq \epsilon \text{ in } \text{poly}(n, \frac{1}{\epsilon})$$

Choose ϵ small enough s.t. $U|\psi\rangle = b'_1 e^{-i\lambda_1 t} |\lambda_1\rangle + \sum_{j=2} c'_j |\lambda_j\rangle$
with $\|b'\|^2 \geq \frac{2}{3}$. ($\epsilon \geq \|U|\psi\rangle - e^{-iHt}|\psi\rangle\| \geq \|b' - b\|$, take $\epsilon = 0.01$ suffices)

Algorithm.

Step 1. Prepare U .

Step 2. Apply phase estimation with precision $\lfloor 2 \log_2 n \rfloor$ on $|\psi\rangle$ and U
Read output.

By linearity, after phase estimation we reach

$$b'_1 |ps(1)\rangle |\psi\rangle + \sum_{j=2} c'_j |ps(j)\rangle |\psi\rangle$$

where $|ps(j)\rangle$ is $\lfloor 2 \log_2 n \rfloor$ bit value of j th eigenvector

Measurement yields success probability $\|b'\|^2 \geq \frac{2}{3}$.

Total cost is $\text{poly}(n)$.

4 Testing graph properties

(a) Left part has $2^{2n+1} - 1$ vertices
 right part has $2^{n+2} - 2$ vertices

$$P = \frac{2^{2n+1} - 1}{2^{2n+1} + 2^{n+2} - 2} = 1 - O(2^{-n})$$

(b) Algorithm.

Step 1. random choose a vertex

Step 2. Do random walk until reach S or a leaf.

Step 3. If reaches a leaf, reach from level 0 to level 1.

For $k=1, \dots, 2n-1$ do:

Determine which one of two edges is towards S
 by walking k steps and see whether reaches a leaf.
 Reach from level k to level k+1.

End for

Analysis: Step 1 has high probability of reaching the left side;
 Step 2 can be seen as random walk on $\mathbb{Z} \cap [1, 2^n]$



Expected hitting time of 0 (or $2n$) is $O(n)$ as well known in theory of Stochastic Process.

Step 3. Worst case $O(1+2+\dots+2n) = O(n^2)$ time

Step 3 illustration

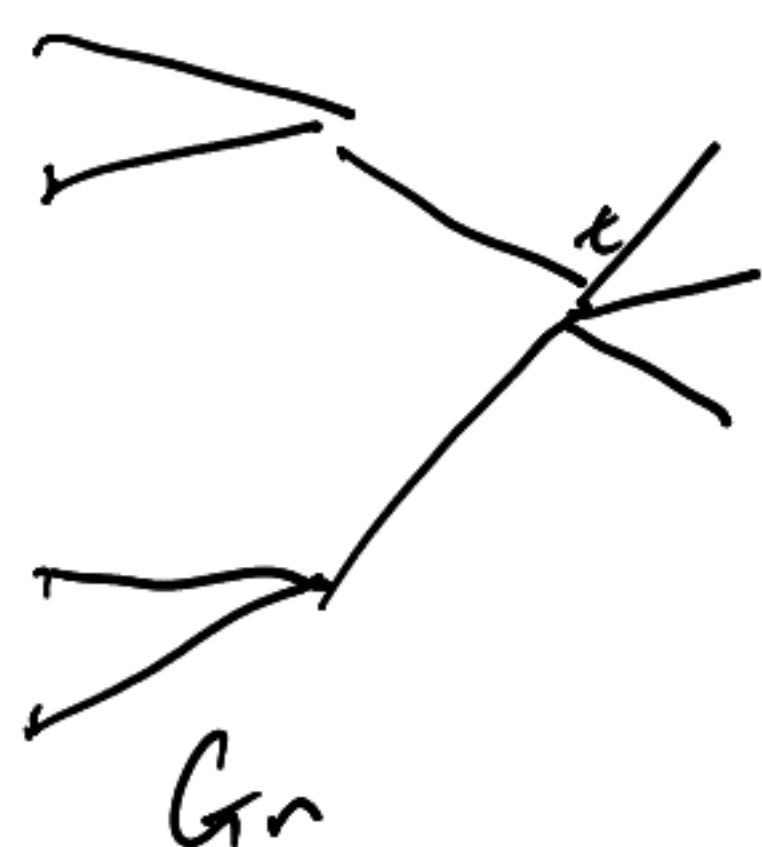


Determine which ? is correct. Searching a complete binary tree from root to leaf takes $O(\text{depth})$ time

Conclusion
 $O(n^2)$ with high probability

(c) Consider G_n be adding 3 edges to t in above graph.

G_n has $2^{\Theta(n)}$ edges



Why classical algorithm must take $2^{\Omega(n)}$ queries:

All vertices being equal, there is high probability to fall in the left part of graph. By to travel to t from some left graph vertex v must travel $v \rightarrow s$ and $s \rightarrow t$ (as each step only knows local information)

But $s \rightarrow t$ is equivalent to the $s-t$ connectivity problem and needs at least $2^{\Omega(n)}$ queries.

Why quantum algorithm works in $\text{poly}(n)$:

Step 1 Pick a random vertex

Step 2 Use (b) to reach s in $O(n^2)$ queries

Step 3 For the 3 other vertex of s , use $O(3 \cdot 2n) = O(n)$

queries to determine whether it points left or right.

(random walk without going back to previous vertex.

reach a leaf exactly in $2n$ steps without reaching t or s if and only if points left)

Step 4. Delete two edges in s pointing left

Use continuous quantum walk to reach t in $\text{poly}(n)$ queries

Total $\text{poly}(n)$ queries with high probability. #