



UNIVERSITÀ DEGLI STUDI DI MILANO

Quantum walks with time-dependent Hamiltonians and their application to the search problem on graphs

Relatore: **Prof. Matteo G. A. Paris**
Correlatore: **Prof. Stefano Olivares**
Correlatrice: **Dott.ssa Claudia Benedetti**

Candidato: **Matteo Garbellini**

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Unstructured search problem

- N items $\{1, 2, \dots, N\}$
- One "marked item" w represents the solution
- Query: "is $w = x$?" is a black box function

$$f(x) = \begin{cases} 0 & x \neq w \\ 1 & x = w \end{cases}$$

- **Classical:** $O(N)$

Grover's quantum search

- Takes advantage of the **superposition** of states
- Algorithm consists in repeated application of the **Grover's operator** $U_G = U_\psi U_w$
- **Oracle** U_w : a black box that **marks** the solution with a phase

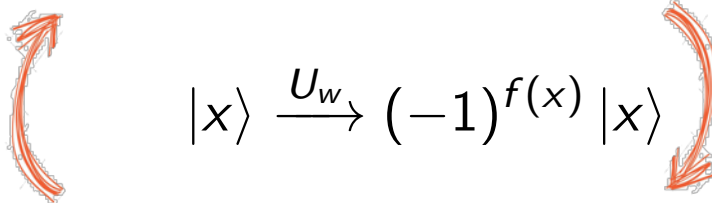
- **Quantum Grover**: $O(\sqrt{N})$

Search procedure

- Preparation of the superposition state

$$|\psi\rangle = \frac{1}{\sqrt{N}} \sum_x |x\rangle$$

- Oracle application


$$|x\rangle \xrightarrow{U_w} (-1)^{f(x)} |x\rangle$$

- Application of U_ψ

$$2|\psi\rangle\langle\psi| - I$$

- Readout

Spatial search by quantum walks

- N elements distributed in space, described by **N -vertex graph**
- Quantum walk dynamics determined by the **Laplacian** L of the graph
- Marked state is **identified** by an **oracle** $H_w = -|w\rangle\langle w|$

$$H = -\gamma L - |w\rangle\langle w|$$

- **Complete graph**: $O(\sqrt{N})$

Search procedure

- Start in state $|s\rangle = \frac{1}{\sqrt{N}} \sum_j |j\rangle$
- Schrodinger evolve for time T using the Hamiltonian H
- Measure the state
- Goal: choose γ , T so that the probability $|\langle w|e^{-iHT}|s\rangle|^2$ is as close to 1 for the smallest T .

A. Child and J. Goldstone, Spatial search by quantum walk. *Physical Review A* (2004)

Search by adiabatic evolution

- The **adiabatic theorem** ensures that under certain conditions if a system evolves slow enough, it remains in its ground state
- It is used to solve computational problem via a **time-dependent evolution**:

$$H(s) = (1-s)H_i + sH_f$$

Global adiabatic search

- Adiabatic theorem is applied globally
- Linear $s(t)$
- Time scaling: $O(N)$

Local adiabatic search

- Adiabatic theorem is applied locally
- Non-linear $s(t)$
- Time scaling: $O(\sqrt{N})$

J. Roland and N. Cerf, Quantum search by local adiabatic evolution. *Physical Review A*

Impossibility of adiabatic quantum walk search

- Both approaches individually perform well, what happens if we combine them?
- Wong *et. al* (2016) showed that it is **impossible** to construct an **adiabatic-quantum walk** search algorithm (for the complete graph)
- It would need a stronger structure than the Grover's oracle

It leaves space for a time-dependent approach **inspired** by the adiabatic evolution, but **free** from the **constraints** of the adiabatic theorem

T. Wong and D. Meyer, Irreconcilable difference between quantum walks and adiabatic quantum computing. *Physical Review A* (2016)

Time-dependent Hamiltonian

- We consider a time-dependent Hamiltonian **inspired** by the adiabatic evolution

$$H(s) = (1 - s)L - s\gamma |w\rangle \langle w|$$

where L is the **Laplacian** of the graph, s is the **interpolating schedule** and $|w\rangle \langle w|$ is the **oracle** Hamiltonian

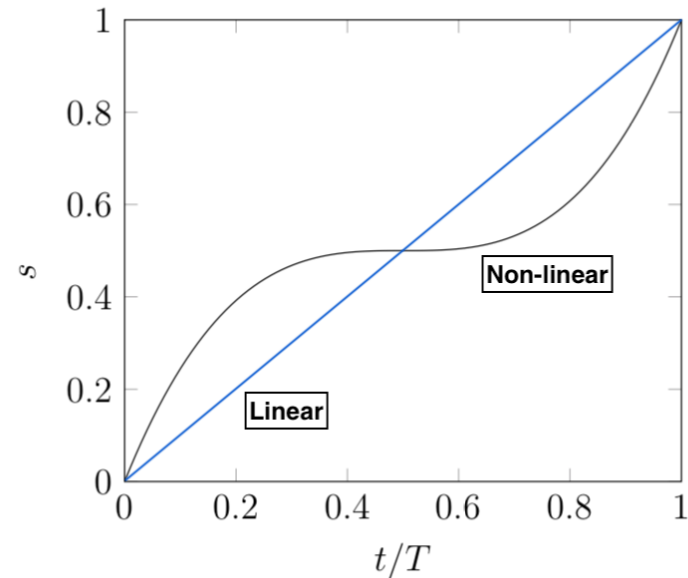
- The evolution of the state is determined by solving the schroedinger equation

$$i \frac{d}{dt} |\psi(t)\rangle = H |\psi(t)\rangle$$

with the necessary boundary conditions.

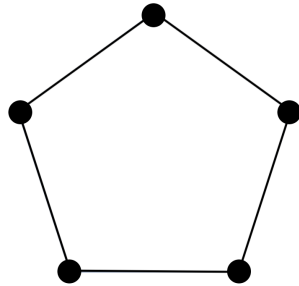
Interpolating schedule and multiple runs for one search

- We considered **linear** $s_L(t)$ and **non-linear** $s_{NL}(t)$ interpolating schedules



- If the search is **imperfect** (solution found with $p < 1$) we consider the possibility of repeating the search **multiple times**

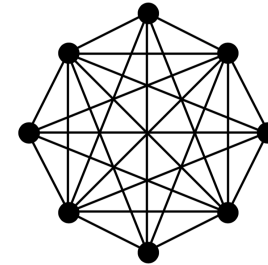
Selected graphs: cycle and complete



Cycle graph $C_y(N)$

Worst case scenario

- Search problem is **not solved** with the standard quantum walks approach



Complete graph $C(N)$

Best case scenario

- Search problem **solved** with standard quantum walks approach
- The equivalent *unstructured search* is **solved** with the local adiabatic evolution

Results characterization: search, localization and robustness

Search

Finding of the solution with high probability for the **smallest** T possible

Localization

Finding of the solution with high probability **without** the need to **minimize** the time

A description of the **localization** is necessary given the **adiabatic nature** of the algorithm

Robustness R_γ and R_T

Quantifies the variation of the probability due to variation of γ and T

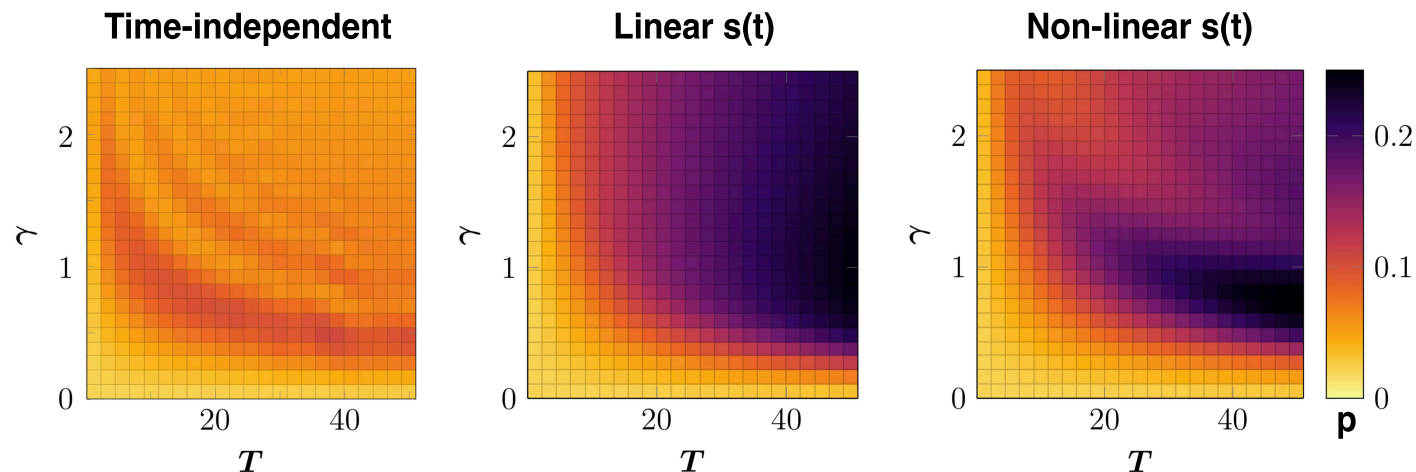
$$R^\pm = p(T, \gamma) - p(T, \gamma \pm \delta)$$

$$R_\gamma = \left[\frac{R^+ + R^-}{2} \right]$$

Cycle graph: probability grid evaluation

We compared the **time-independent** approach and the **time-dependent** one with **linear** and **non-linear** $s(t)$

- Computed the probability over a **grid** for the different combination of T and γ , with graphs up to $N = 71$
- Intuitive way to visualize the results: **probability heatmaps** (in figure $N = 51$)

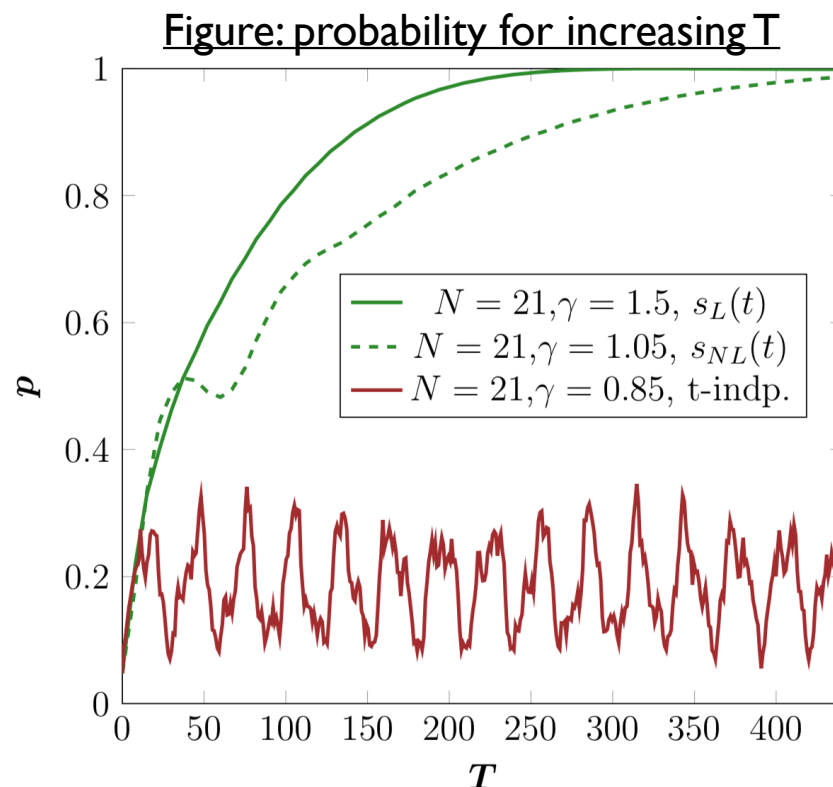


Cycle graph: localization

■ Time-dependent approach shows localization properties

- The time-independent approach (red) does **not** have **localization** properties

- The time-dependent approach (green) is able to achieve $p = 1$, although for **large** T ($\approx N^2$)



- Time-dependent approach produces high probability ($p < 1$) in much less time than for $p = 1$ (e.g. for $s_L(t)$ $p = 1$ at $T = 300$, $p = 0.9$ at $T = 150$)

Cycle graph: search

■ Multiple runs search with constraint on T

- We consider the multiple runs for one search approach
 - (*) Time-dependent is not optimized on T
 - (*) Time-independent is an imperfect search
- We introduce a **new quantity** τ in order to compare the two approaches

$$\tau = \min \left(\frac{T}{p} \right)_{T > T_{\min}, \gamma}$$

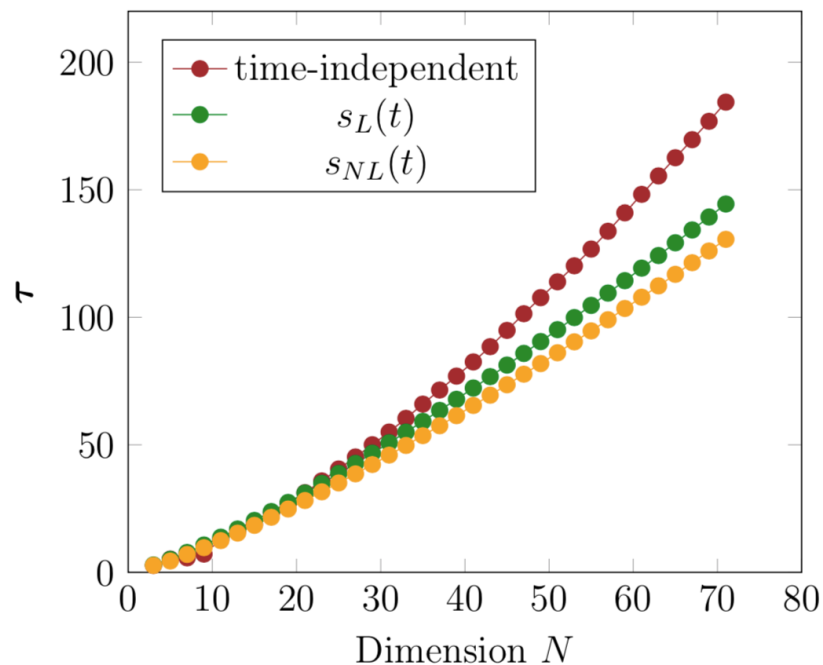
- T_{\min} is a constraint on the minimum time
- $T_{\min} = \frac{\pi}{2} \sqrt{N}$

Cycle graph: search (2)

■ Time-dependent performs better than time-independent

- Similar performance for small N (< 35)
- Time-dependent approach wins for large N
- Non-linear $s(t)$ performs slightly better than the linear one

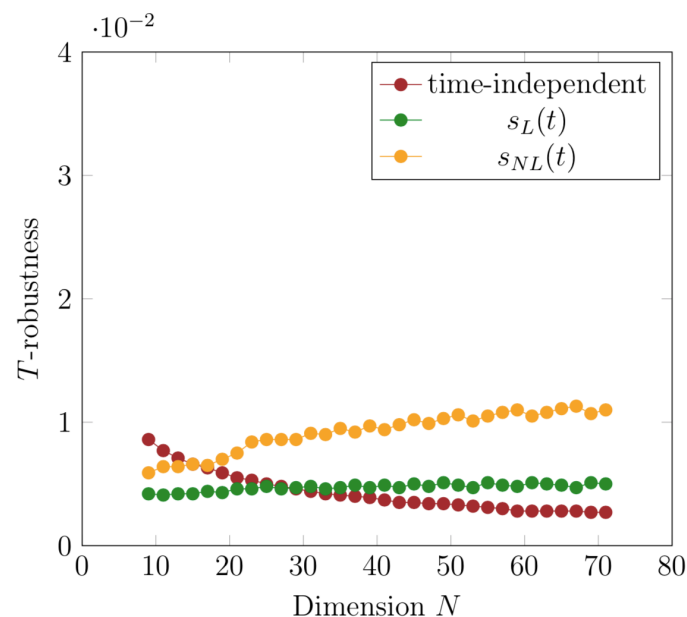
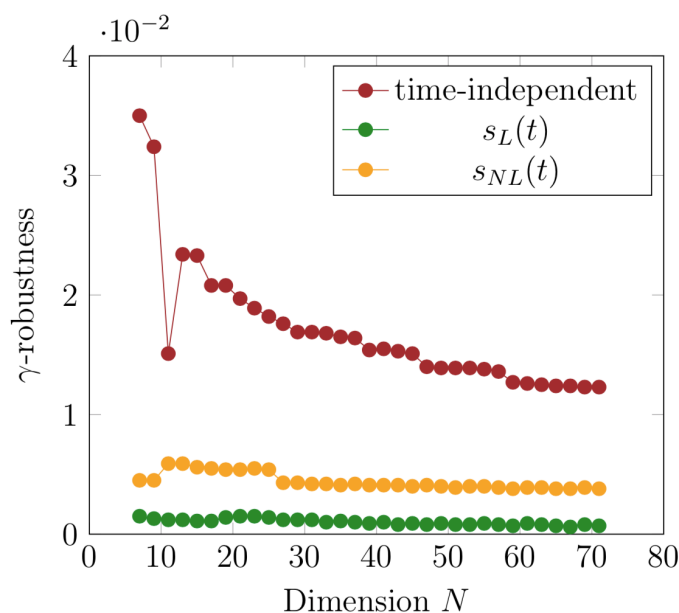
Figure: tau for increasing graph dimension



Cycle graph: robustness

■ Time-dependent approach with linear $s(t)$ is the most robust

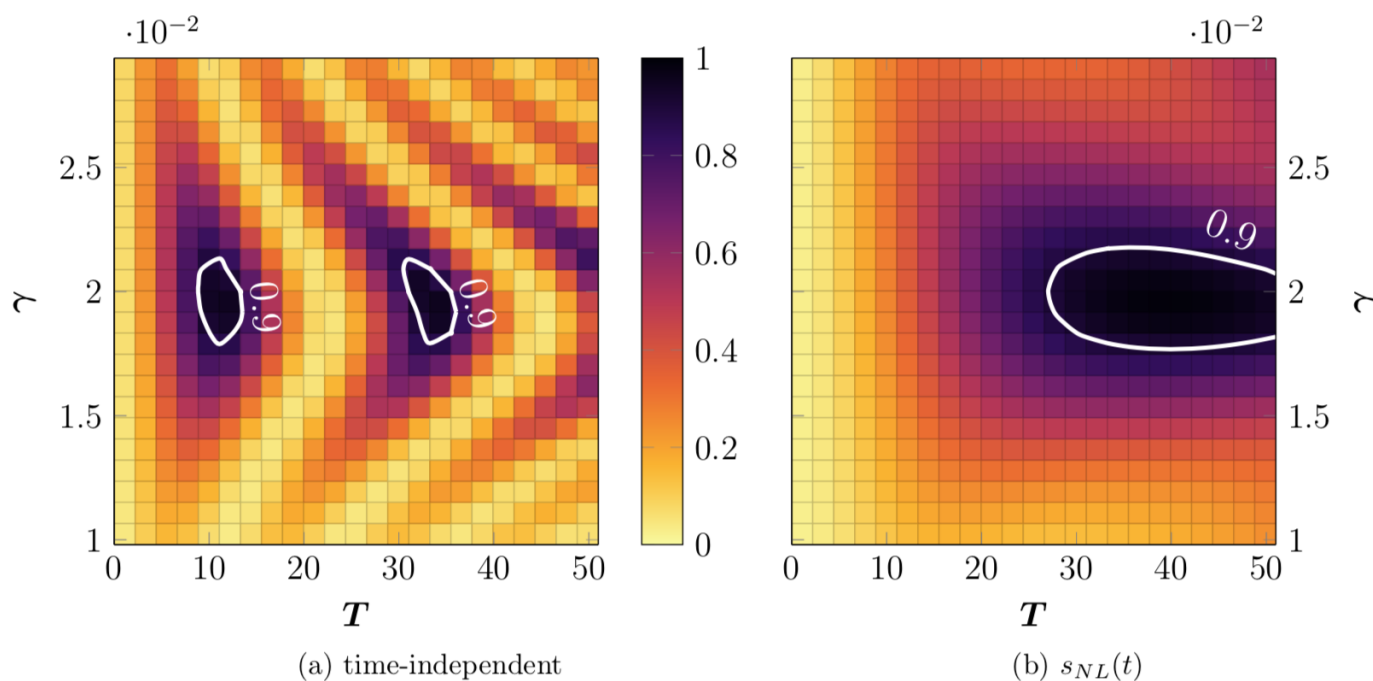
- Robustness is evaluated for a T, γ variation of 2 **grid blocks**
- **γ -Robustness**: time-dependent is best, with linear $s(t)$ the most robust
- **T -Robustness**: surprisingly the time-independent is more robust for large N , although the values are very similar



Complete graph: probability and qualitative robustness

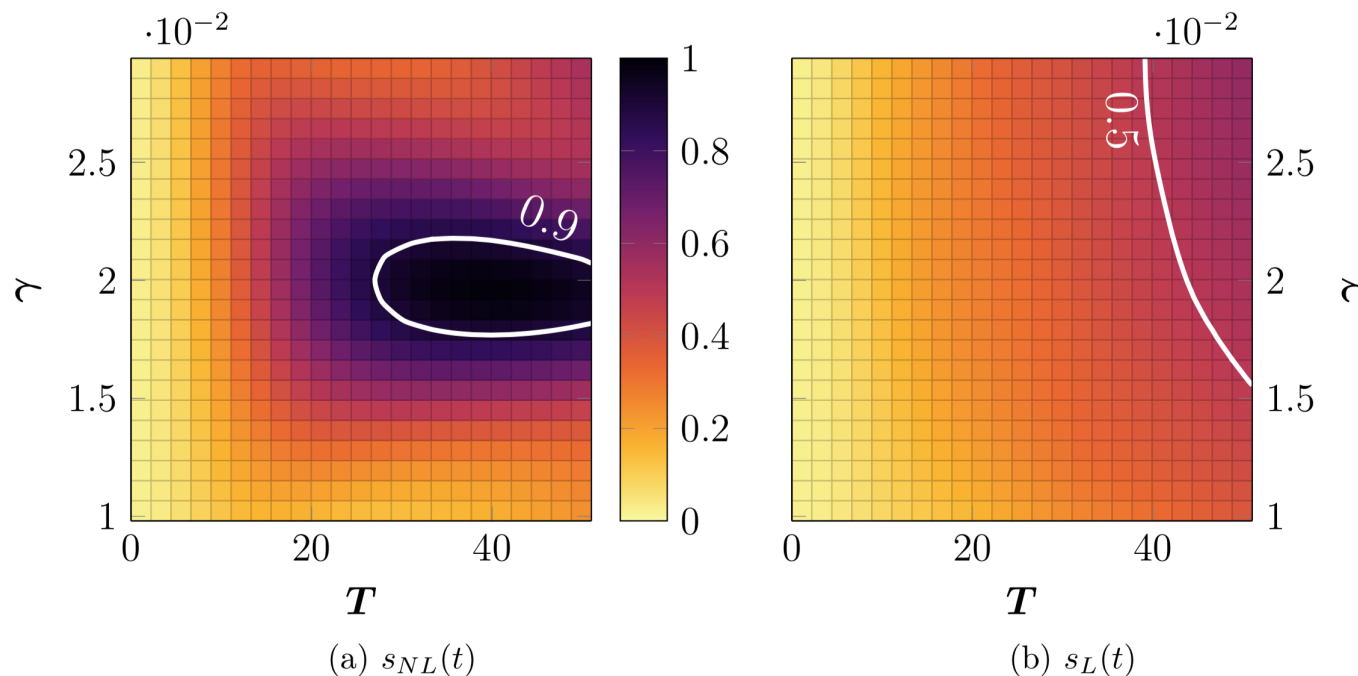
■ Time-dependent is more robust but with much worse time scaling

- For completeness we study the complete graph
- We do not expect to be able to improve an already optimal approach
- We might get some insights on the performance of our algorithm
- We study the **qualitative robustness** (in figure $N = 51$)



The choice of $s(t)$ is crucial

- As already mentioned the choice of $s(t)$ is crucial
- Improvements on the performance come from the choice of the optimal $s(t)$
- Linear $s(t)$ performs **worse** than **classical** algorithms (in figure $N = 51$)



What have we learned?

Results from applying the time-dependent Hamiltonian algorithm

Cycle graph

- Shows localization properties
- Is more robust
- Better search performance overall
- Non-linear $s(t)$ performs better but is less robust

Complete graph

- Is (qualitatively) more robust
- Performs much worse than Grover
- Performs better than classical
- Shows that $s(t)$ has great impact on the performance

What is next?

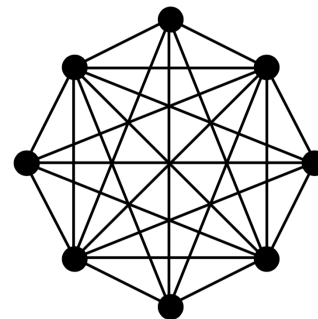
- The application of the algorithm to the complete graph suggests that **improvements** on the performance come from the choice of the **optimal** $s(t)$
- Application to **different graph topologies**

Thanks for your attention

Search by quantum walks

- N elements distributed in space (e.g., a physical database)
- Model: N -vertex graph G
- Dynamics of the quantum walk on the graph is determined by the Laplacian L of the graph

$$H = -\gamma L$$



Adjacency matrix:

$$A_{ij} = \begin{cases} 1 & \text{if } (i, j) \in G \\ 0 & \text{otherwise} \end{cases}$$

Diagonal degree matrix:

$$D_{jj} = \deg(j)$$

Laplacian matrix:

$$L = A - D$$