



GUIDED QUANTUM WALK

A New Algorithm between Quantum Walk and Quantum Annealing

7. NOVEMBER 2023 | DENNIS WILLSCH

Guided quantum walk

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Forschungszentrum Jülich, 52425 Jülich, Germany*

²*AIDAS, 52425 Jülich, Germany*

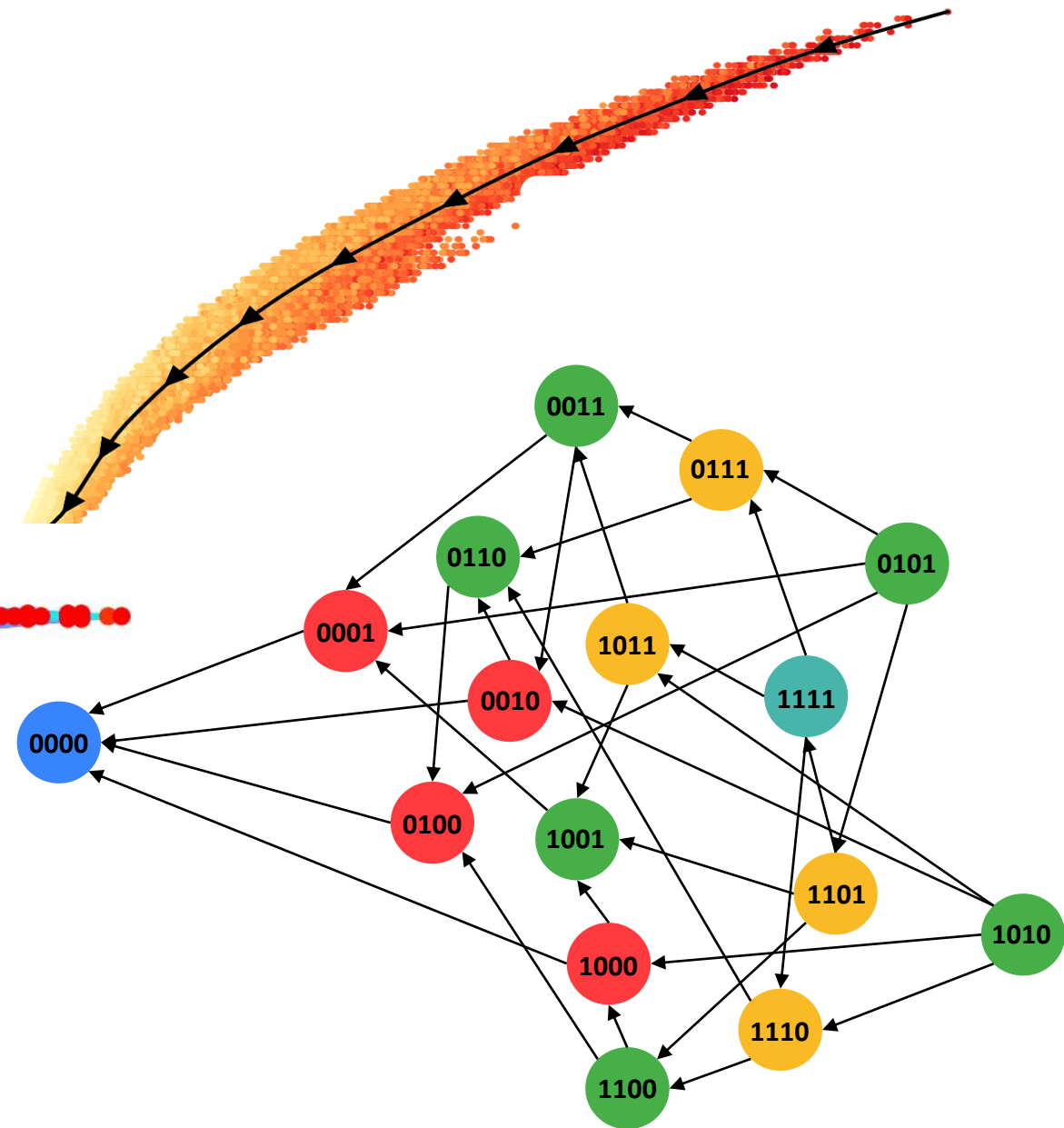
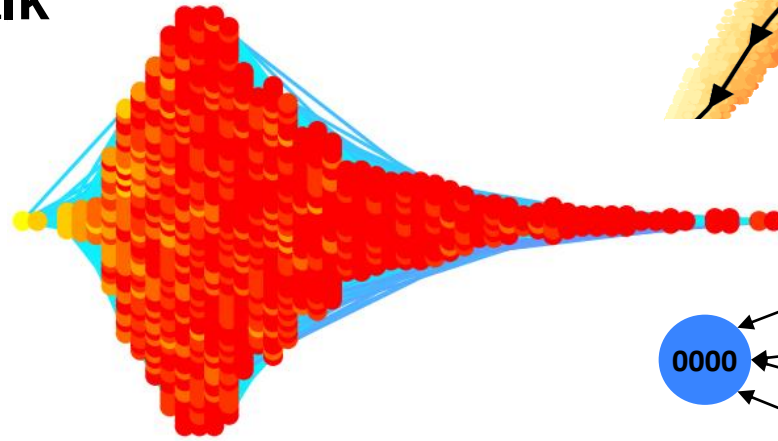
³*RWTH Aachen University, 52056 Aachen, Germany*

arXiv:2308.05418

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Contents

- Local Amplitude Transfer (LAT) Theory
- Guided Quantum Walk
 - Theoretical GQW
 - Practical GQW
- Performance
 - Scaling Analysis
 - Comparison to QA, QW, QAOA, AQA
 - Accounting for Classical Optimization Phase



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Local Amplitude Transfer (LAT) Theory

➤ Hamiltonian:

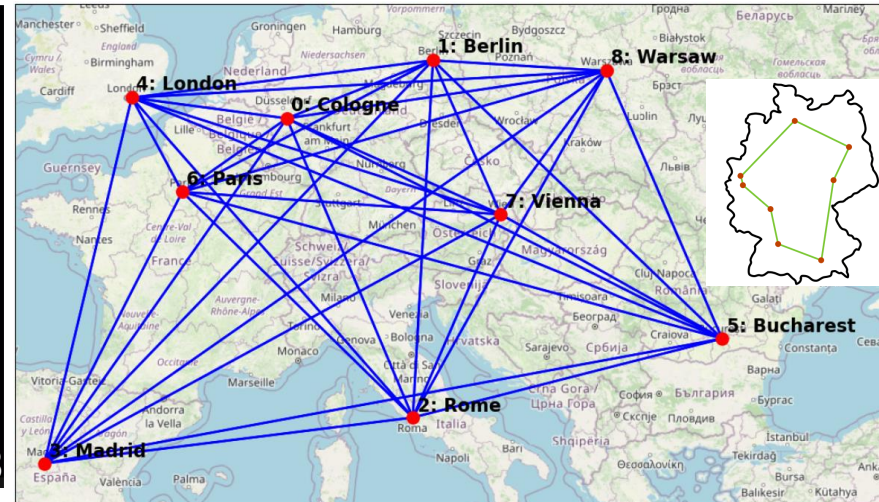
$$H_{GQW}(t) = \underbrace{\Gamma(t) \cdot \left(- \sum_{i=1}^N \sigma_i^x \right)}_{\substack{\text{Driver} \\ \text{Hamiltonian} \\ H_D}} + \underbrace{\left(- \sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)}_{\substack{\text{Cost} \\ \text{Hamiltonian} \\ H_C}}$$

Time-dependent hopping rate

➤ We consider three problems:



Exact Cover (EC)
from Tail Assignment Problems



Traveling Salesperson (TSP)



Garden Optimization (GO)

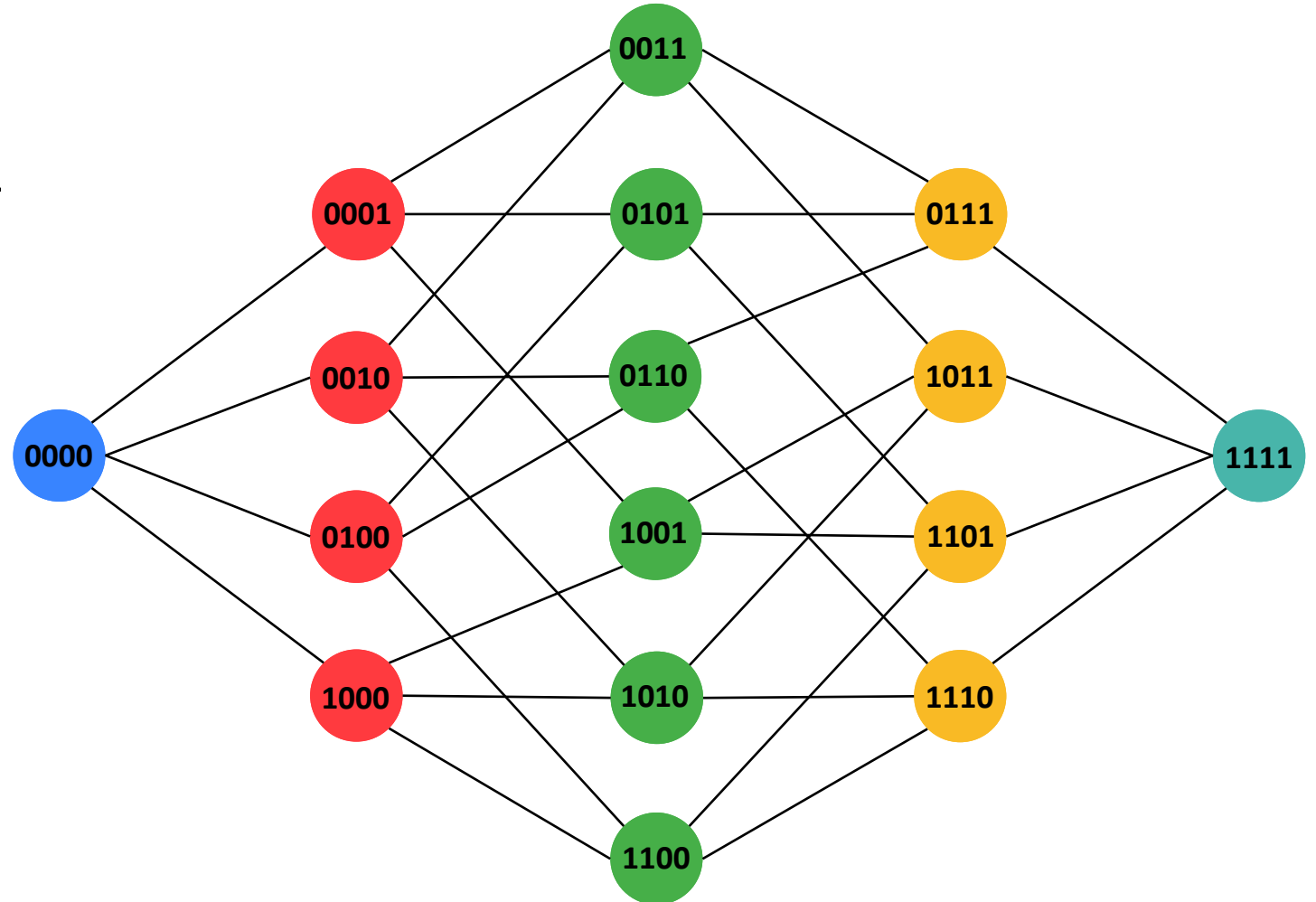
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Local Amplitude Transfer (LAT) Theory

Generic Energy Spectrum of H_C

- States with Hamming distance 1 are coupled by H_D
- Order states by energy E

$$H_{GQW}(t) = \Gamma(t) \cdot \underbrace{\left(-\sum_{i=1}^N \sigma_i^x \right)}_{H_D} + \underbrace{\left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)}_{H_C}$$

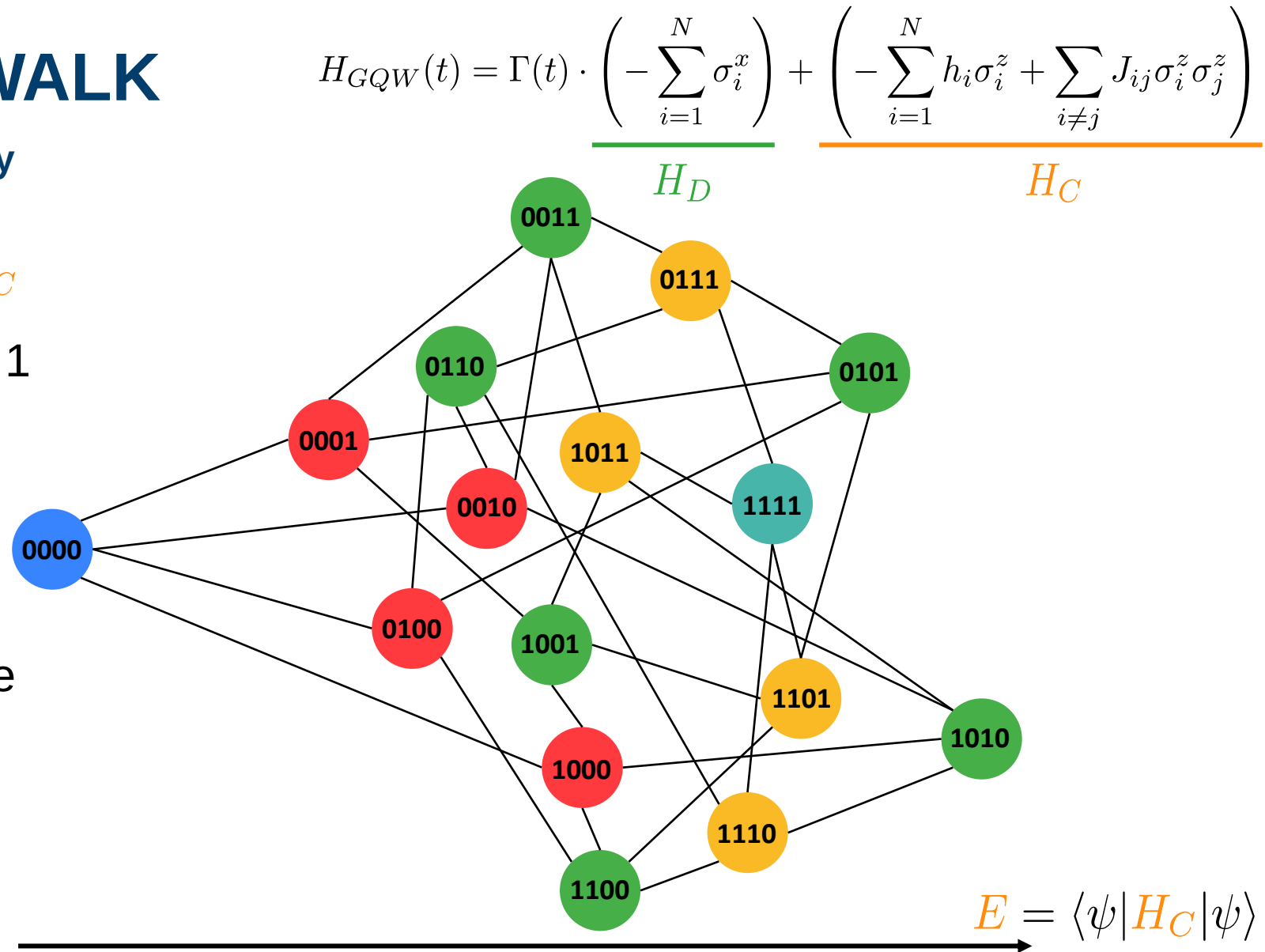


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Local Amplitude Transfer (LAT) Theory

Generic Energy Spectrum of H_C

- States with Hamming distance 1 are coupled by H_D
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- **Goal:** Drive probability amplitude towards ground state



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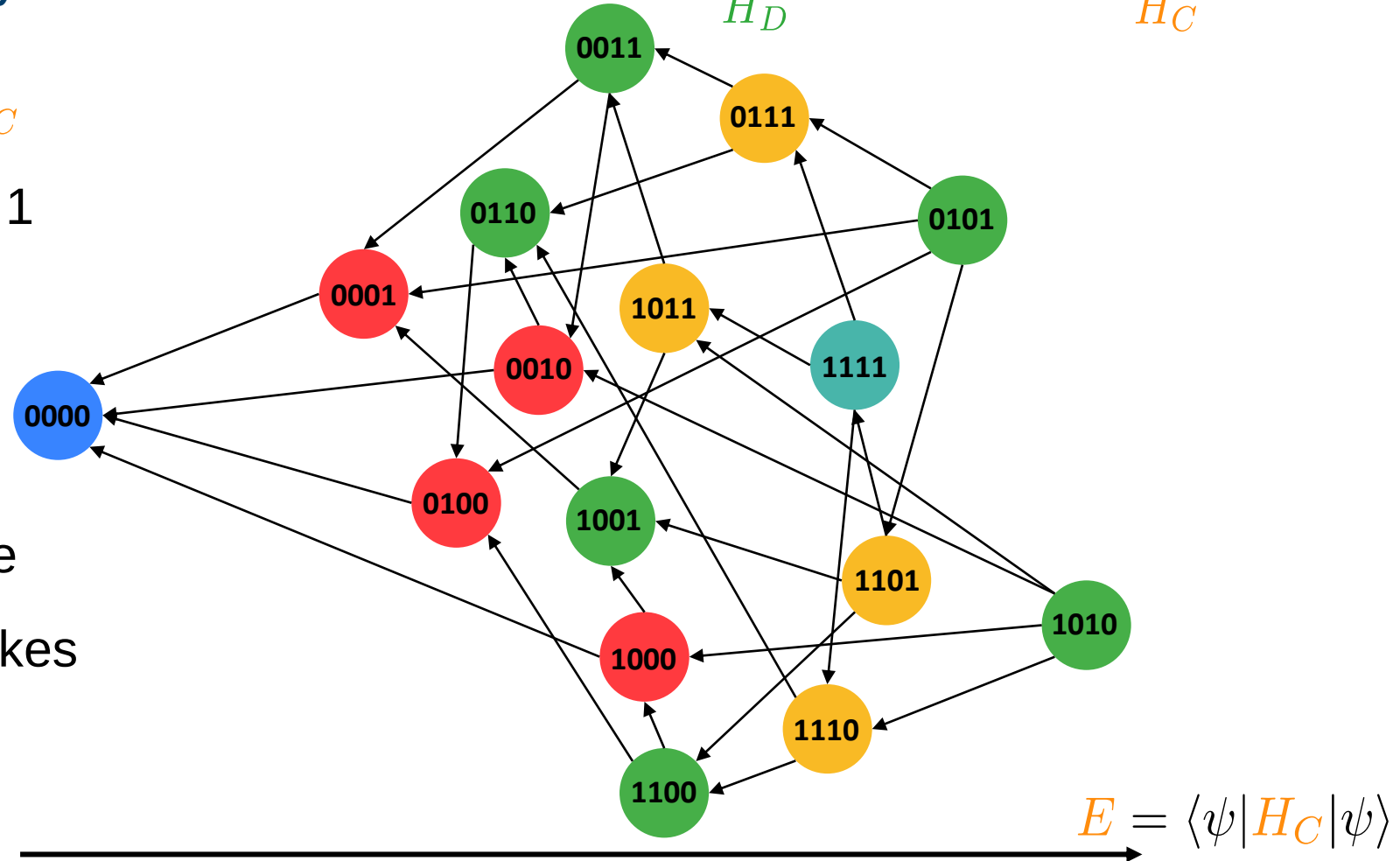
Local Amplitude Transfer (LAT) Theory

Generic Energy Spectrum of H_C

- States with Hamming distance 1 are coupled by H_D
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- **Goal:** Drive probability amplitude towards ground state
- **For large problems:** Graph takes generic “Onion” shape



$$H_{GQW}(t) = \Gamma(t) \cdot \underbrace{\left(-\sum_{i=1}^N \sigma_i^x \right)}_{H_D} + \underbrace{\left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)}_{H_C}$$



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Local Amplitude Transfer (LAT) Theory

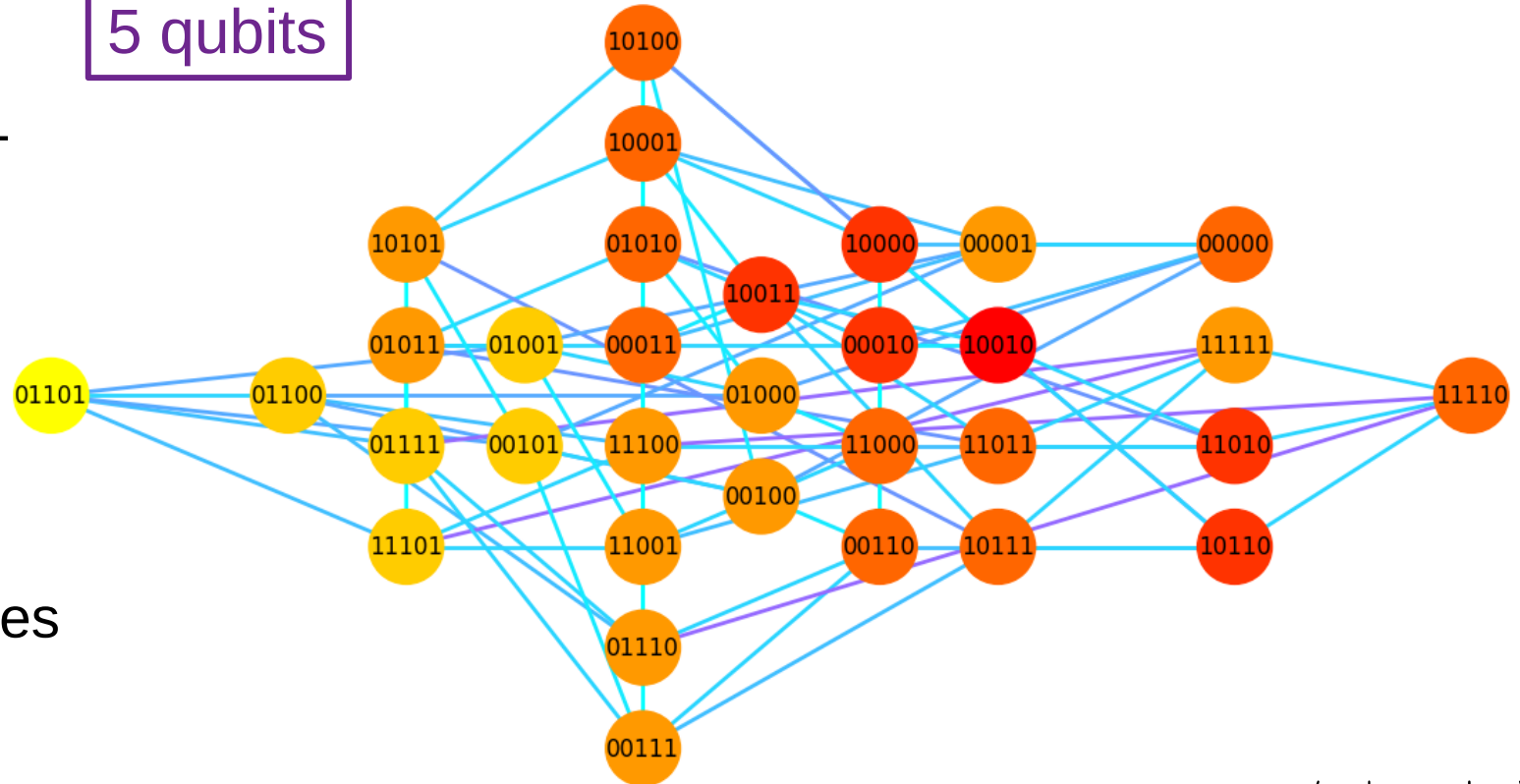
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Generic Energy Spectrum of H_C

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5 qubits



$$E = \langle \psi | H_C | \psi \rangle$$

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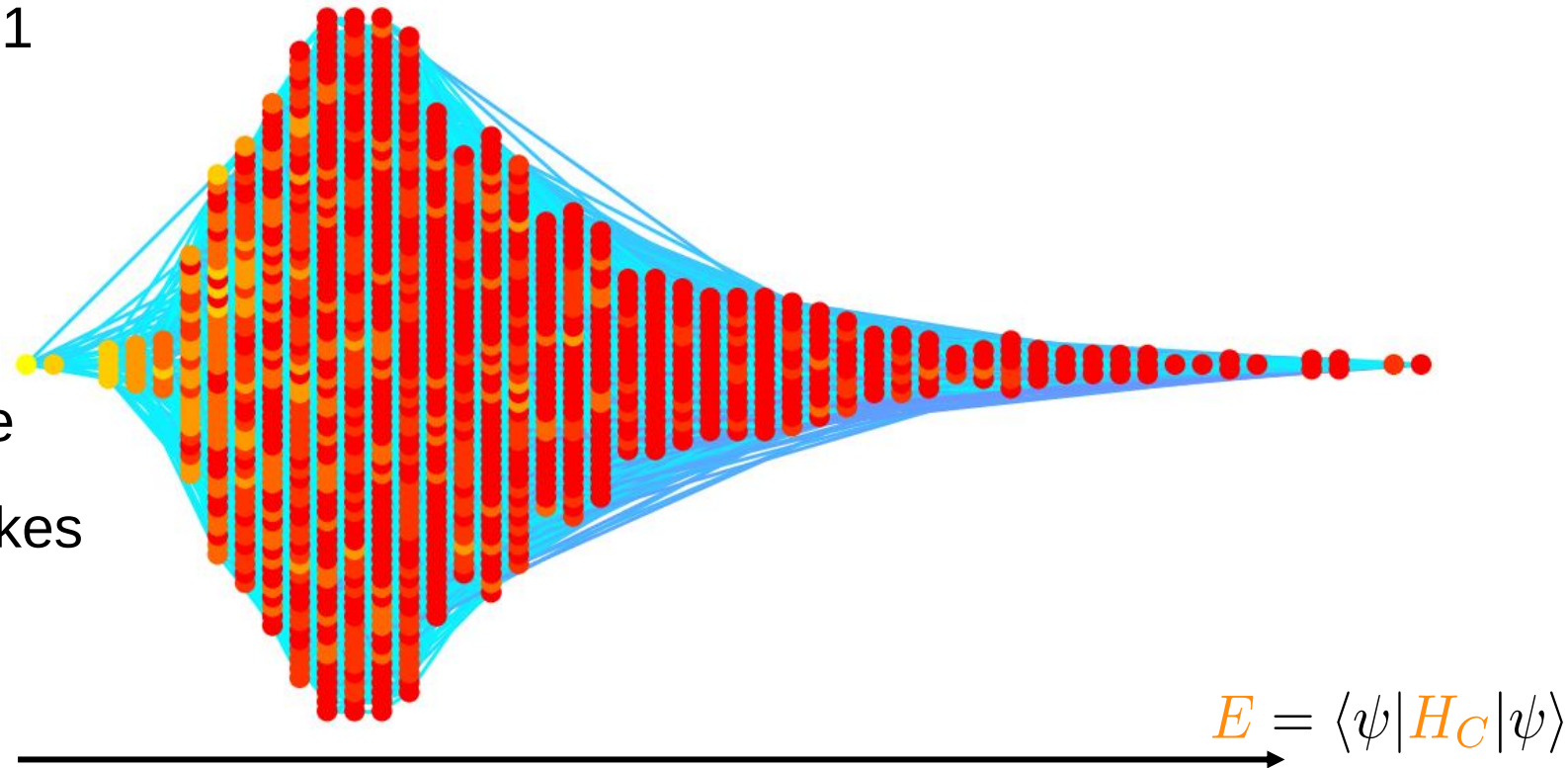
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Generic Energy Spectrum of H_C

10 qubits

- States with Hamming distance 1 are coupled by H_D
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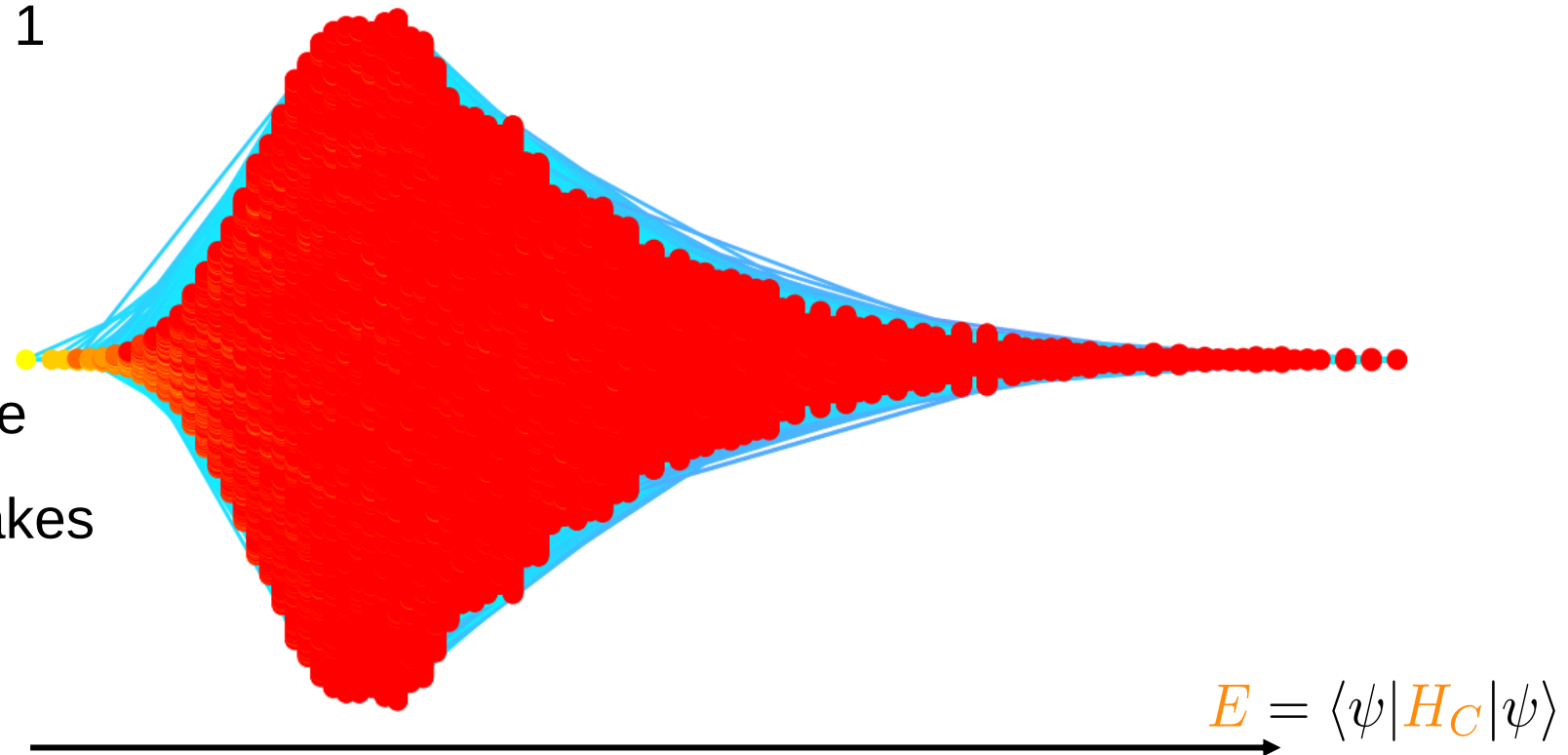
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Generic Energy Spectrum of H_C

15 qubits

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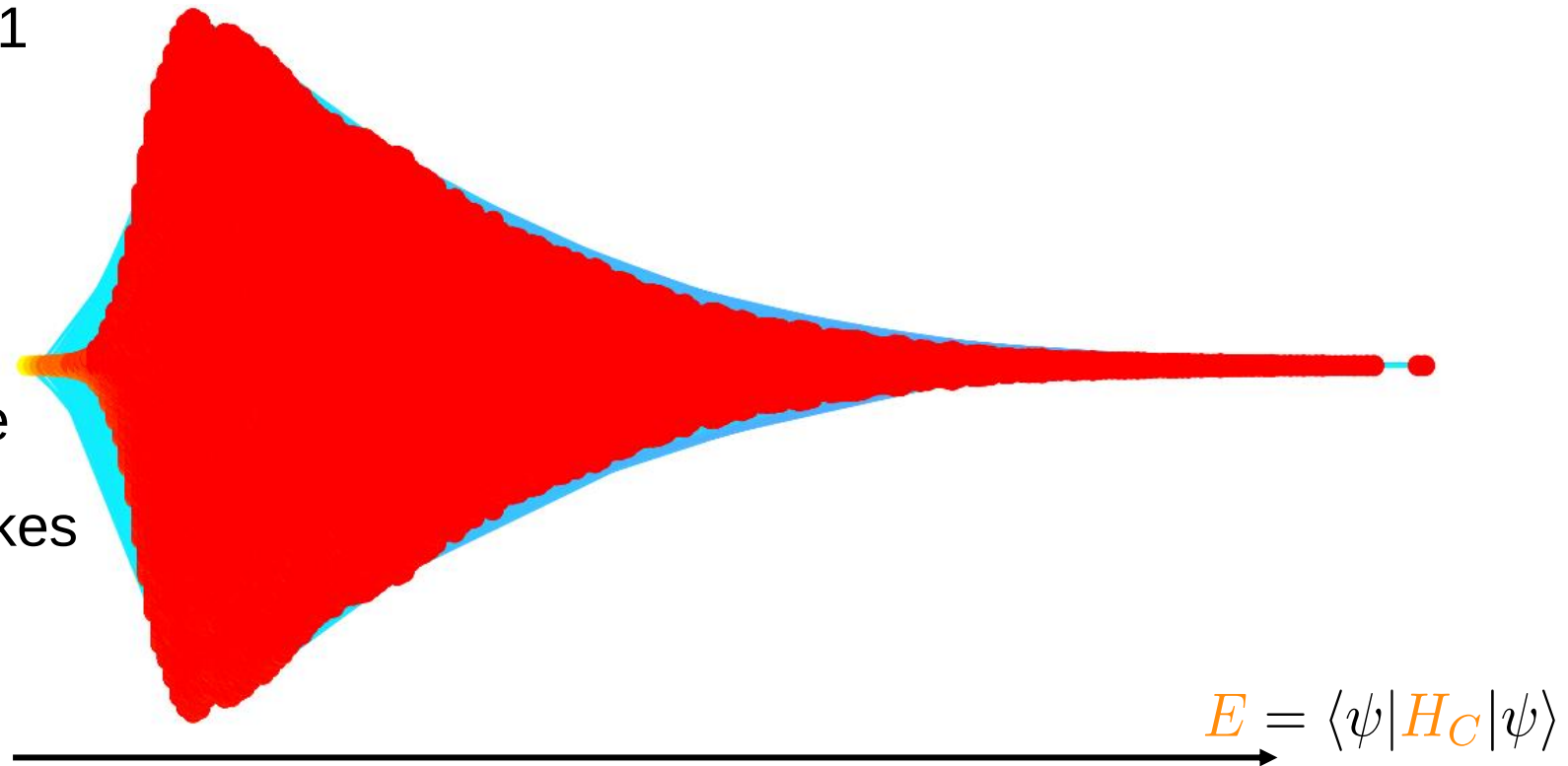
Local Amplitude Transfer (LAT) Theory

$$H_{GQW}(t) = \Gamma(t) \cdot \underbrace{\left(- \sum_{i=1}^N \sigma_i^x \right)}_{H_D} + \underbrace{\left(- \sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)}_{H_C}$$

Generic Energy Spectrum of H_C

18 qubits

- States with Hamming distance 1 are coupled by H_D
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Local Amplitude Transfer (LAT) Theory

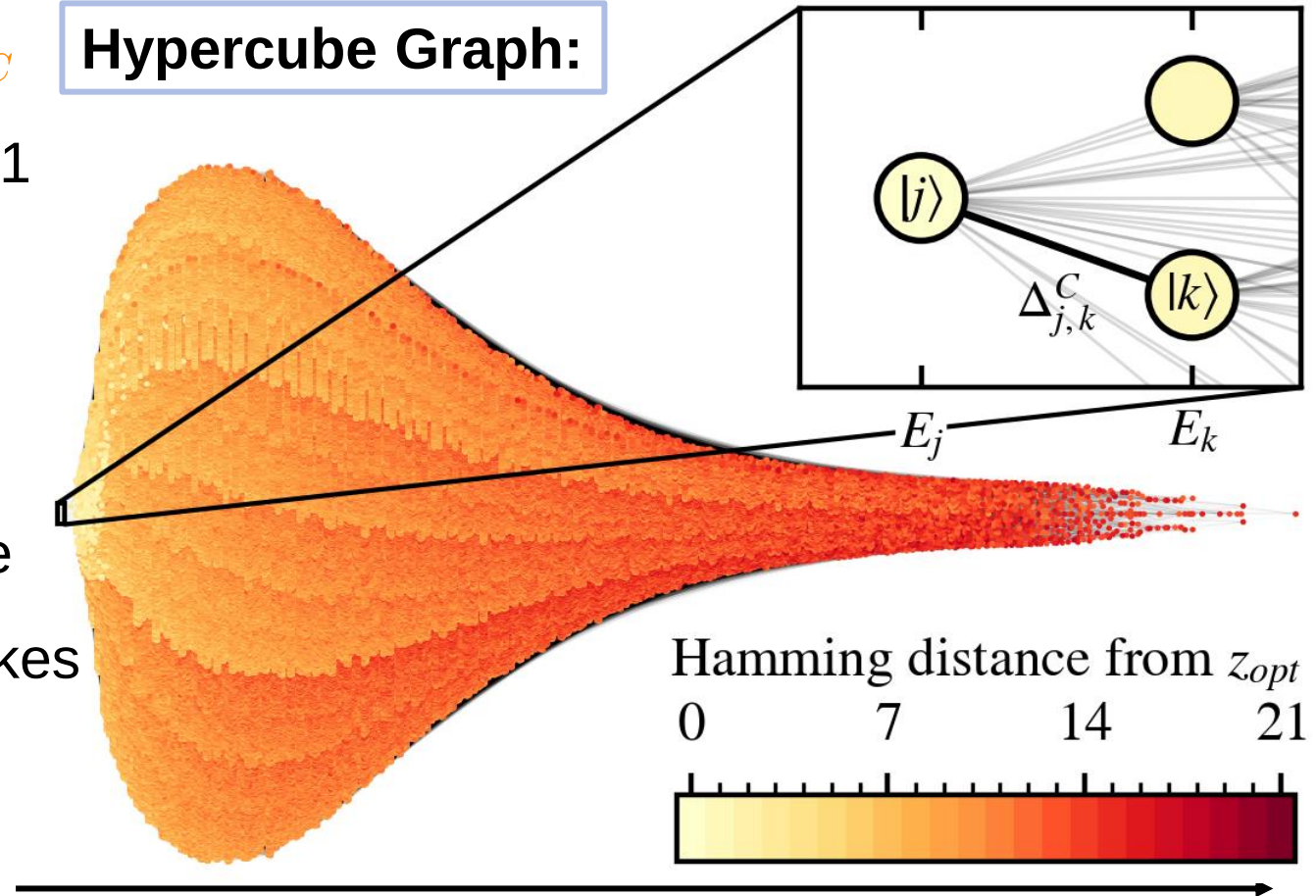
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Generic Energy Spectrum of H_C

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Hypercube Graph:

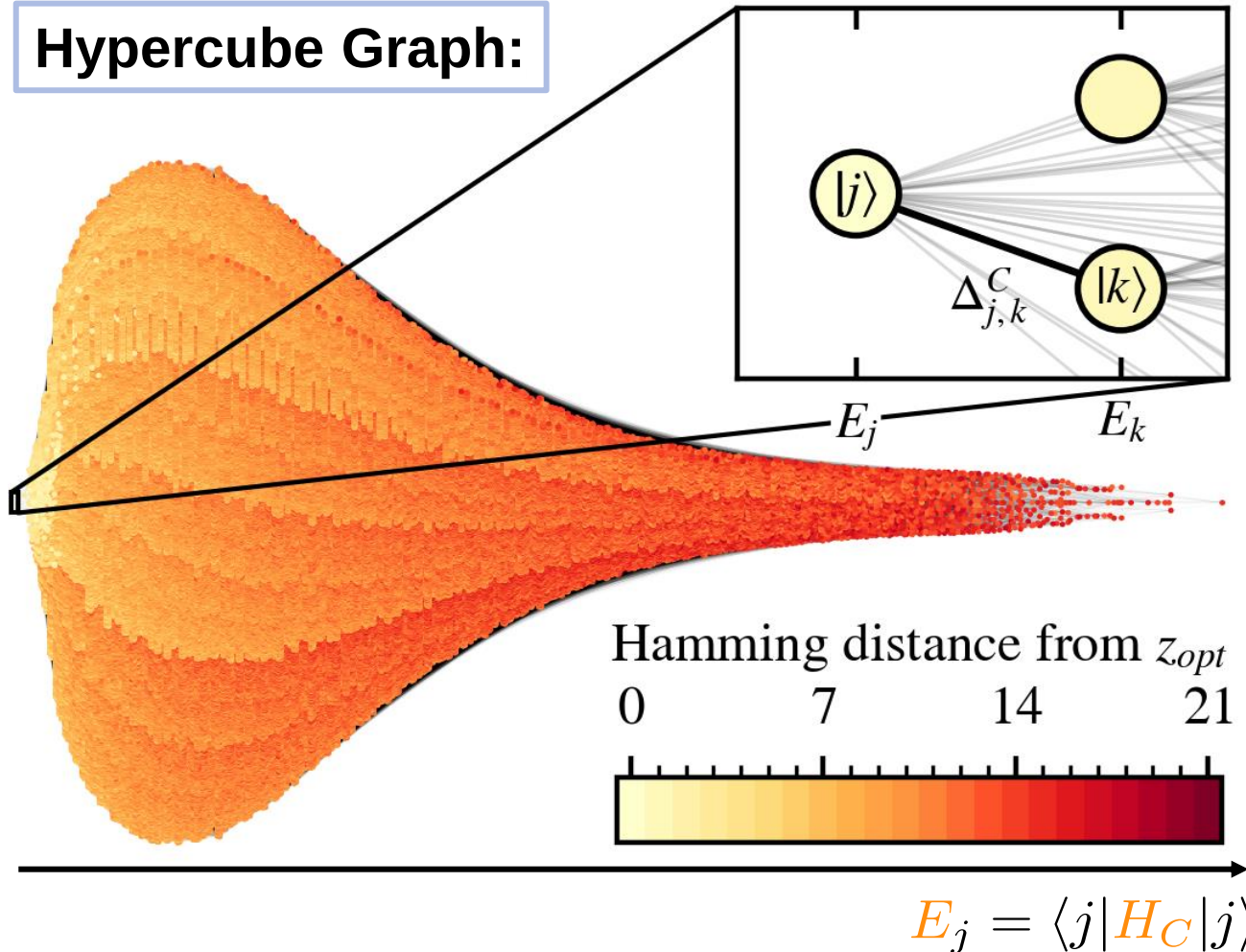


$$E_j = \langle j | H_C | j \rangle$$

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Local Amplitude Transfer (LAT) Theory

Hypercube Graph:



$$H_{GQW}(t) = \Gamma(t) \cdot \underbrace{\left(-\sum_{i=1}^N \sigma_i^x \right)}_{H_D} + \underbrace{\left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)}_{H_C}$$

- Look at **Local** Hamiltonian

$$H^{(j,k)} = \Gamma \cdot H_D^{(j,k)} + H_C^{(j,k)}$$

$$H_D^{(j,k)} = -\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad H_C^{(j,k)} = \begin{pmatrix} E_j & 0 \\ 0 & E_k \end{pmatrix}$$

- Drives **Amplitude Transfer** $|j\rangle \longleftrightarrow |k\rangle$

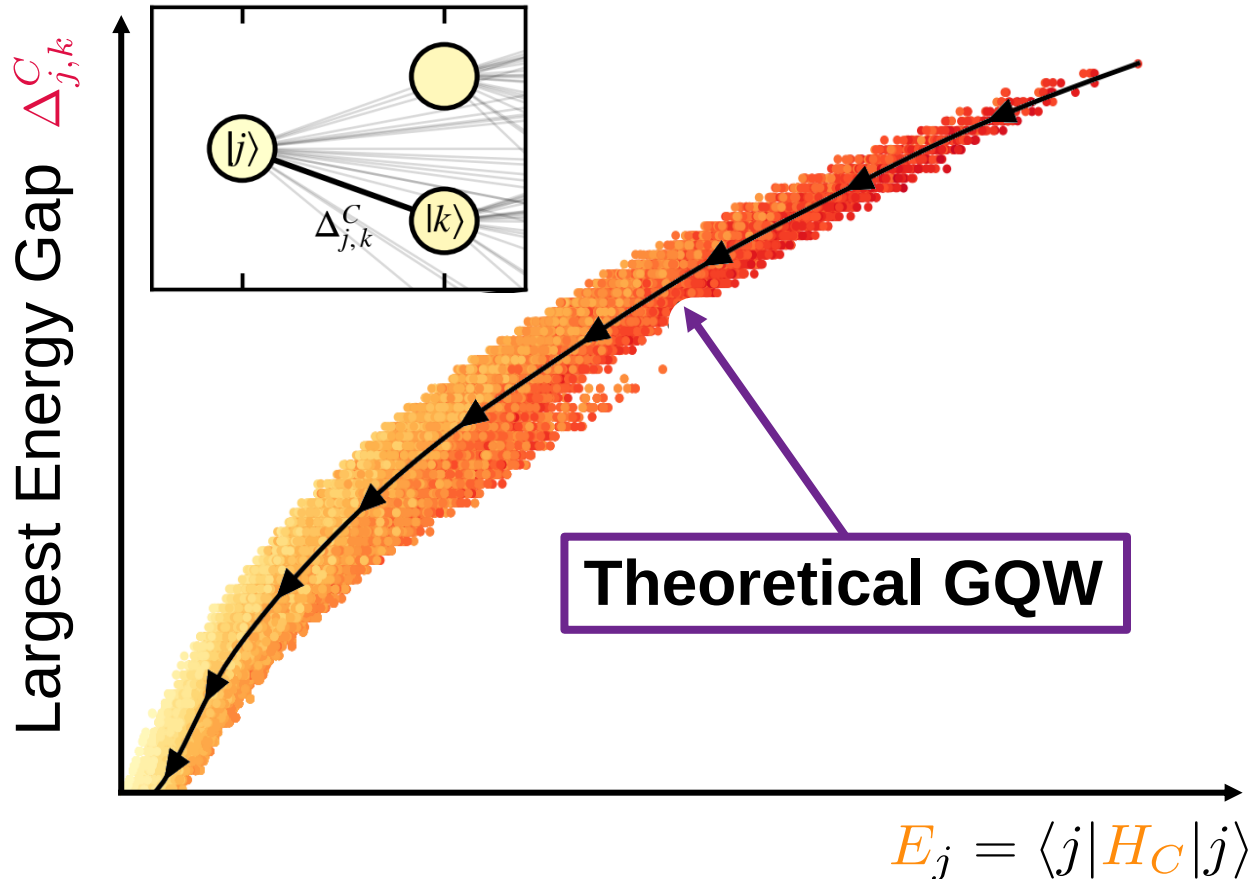
$$P_{\rightarrow |j\rangle}(t) = \frac{1}{2} + \frac{\Gamma \Delta_{j,k}^C}{2\Omega^2} \sin^2 \Omega t$$

- **Local Energy Gaps** $\Delta_{j,k}^C = E_k - E_j$

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Local Amplitude Transfer (LAT) Theory

Hypercube Graph:



$$H_{GQW}(t) = \Gamma(t) \cdot \underbrace{\left(-\sum_{i=1}^N \sigma_i^x \right)}_{H_D} + \underbrace{\left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)}_{H_C}$$

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- **Local Energy Gaps** $\Delta_{j,k}^C = E_k - E_j$

- **Local Rabi oscillations** $\Omega = \sqrt{\Gamma^2 + \left(\frac{\Delta_{j,k}^C}{2} \right)^2}$

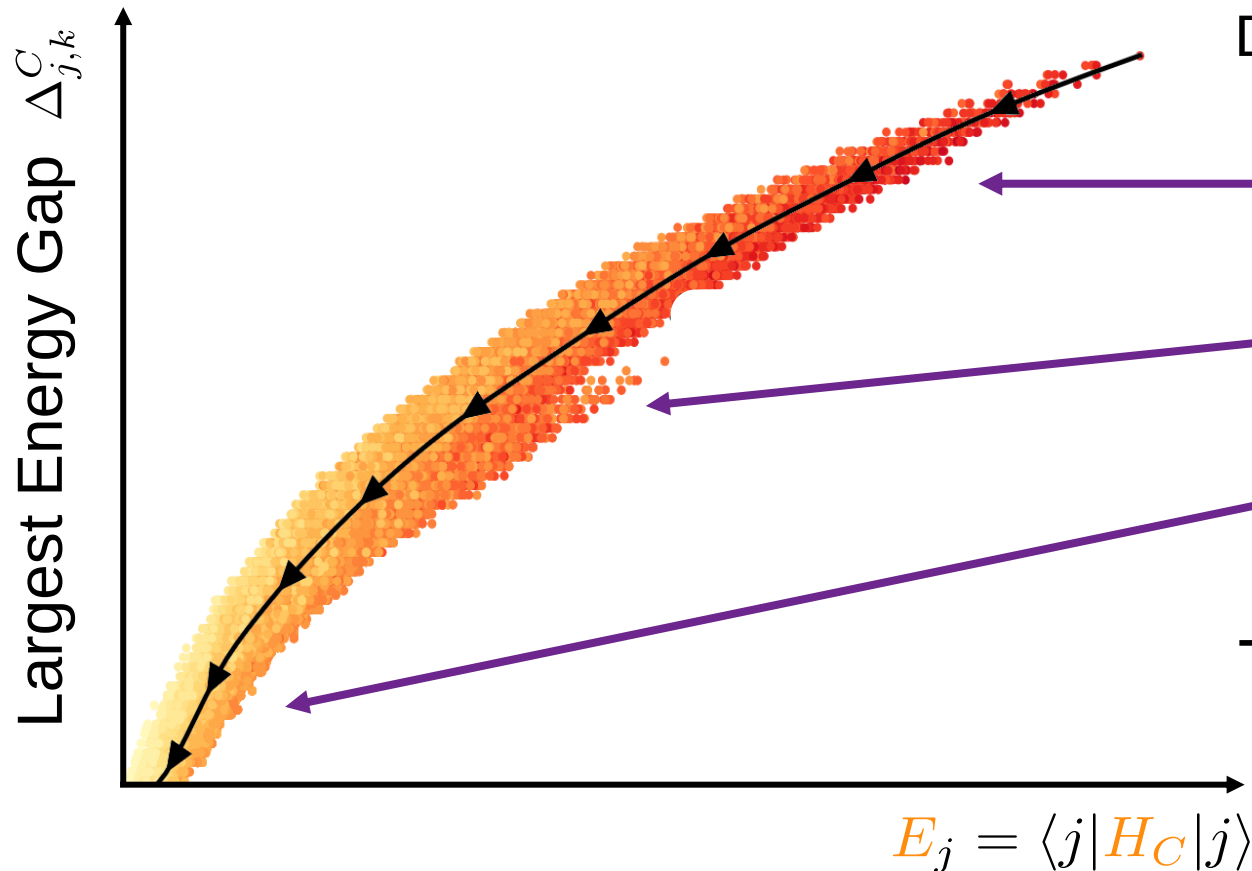
- **LAT Theory**: Optimal Balance at

$$\Gamma \approx \Delta_{j,k}^C / 2$$

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Theoretical GQW

LAT Theory: $\Gamma \approx \Delta_{j,k}^C / 2$



$$H_{GQW}(t) = \Gamma(t) \cdot \left(- \sum_{i=1}^N \sigma_i^x \right) + \left(- \sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)$$

Drive Amplitude Through Hypercube Graph:

Beginning $\Gamma(t) \approx \langle \Delta^C \rangle_{\text{large}} E_j / 2$

Intermediate $\Gamma(t) \approx \langle \Delta^C \rangle_{\text{medium}} E_j / 2$

End $\Gamma(t) \approx \langle \Delta^C \rangle_{\text{small}} E_j / 2$

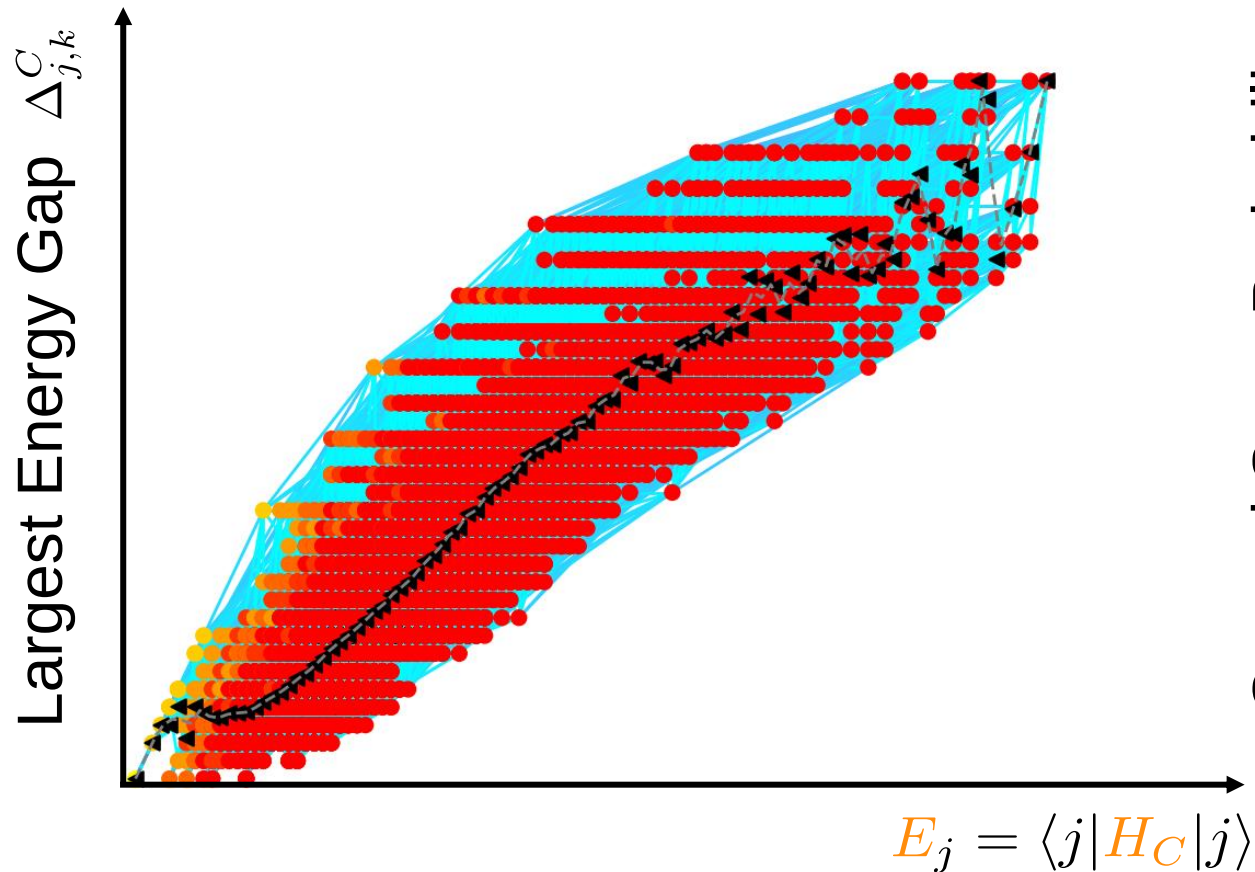
→ Hopping Rate “Schedule”

$$\Gamma(t) \approx \langle \Delta^C \rangle(E(t)) / O(1)$$

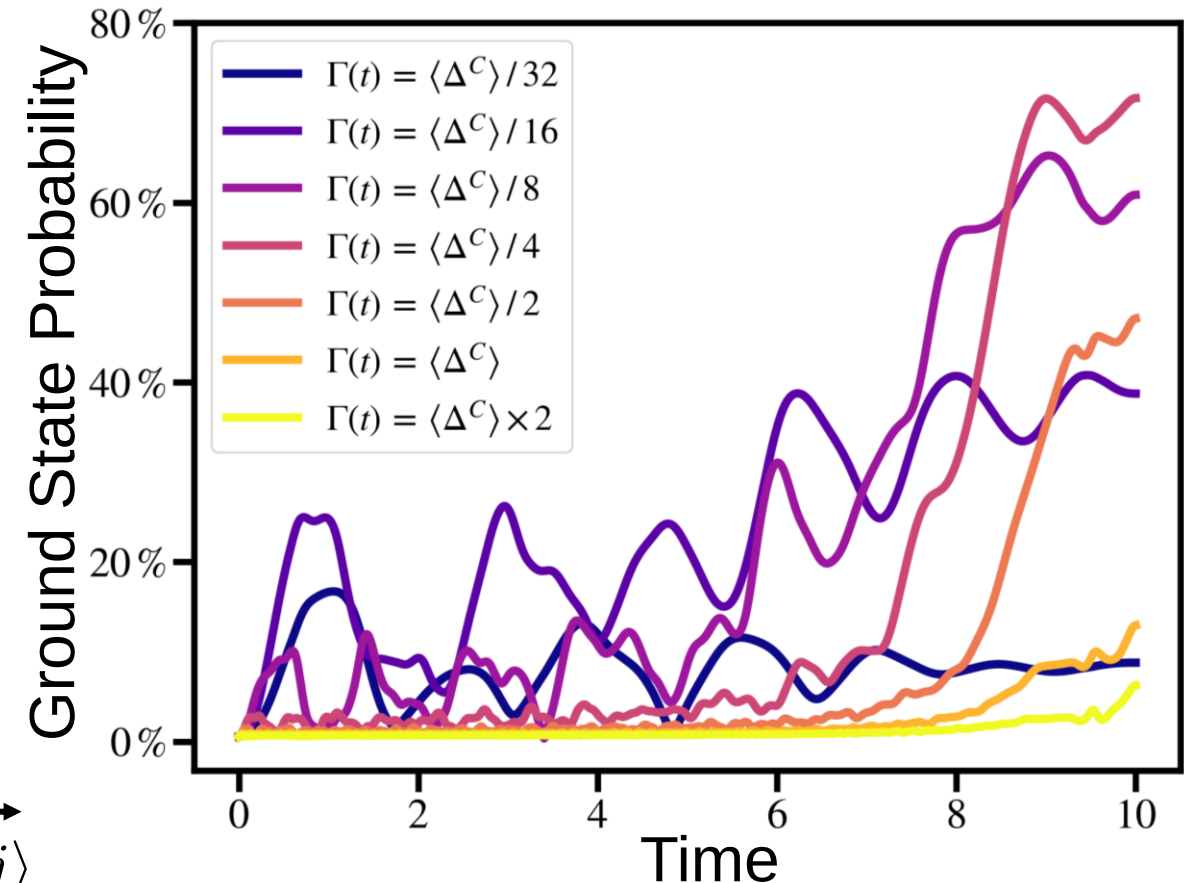
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Theoretical GQW

LAT Theory: $\Gamma(t) \approx \langle \Delta^C \rangle(E(t)) / O(1)$



$$H_{GQW}(t) = \Gamma(t) \cdot \left(-\sum_{i=1}^N \sigma_i^x \right) + \left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)$$



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Practical GQW

➤ In practice: $\langle \Delta^C \rangle(E(t))$ is not known

➤ Observation:

➤ Generic energy spectrum

➤ $\Gamma(t)$ takes generic shape

“Typical” Quantum Annealing Schedule!

➤ Deviations can be modelled with

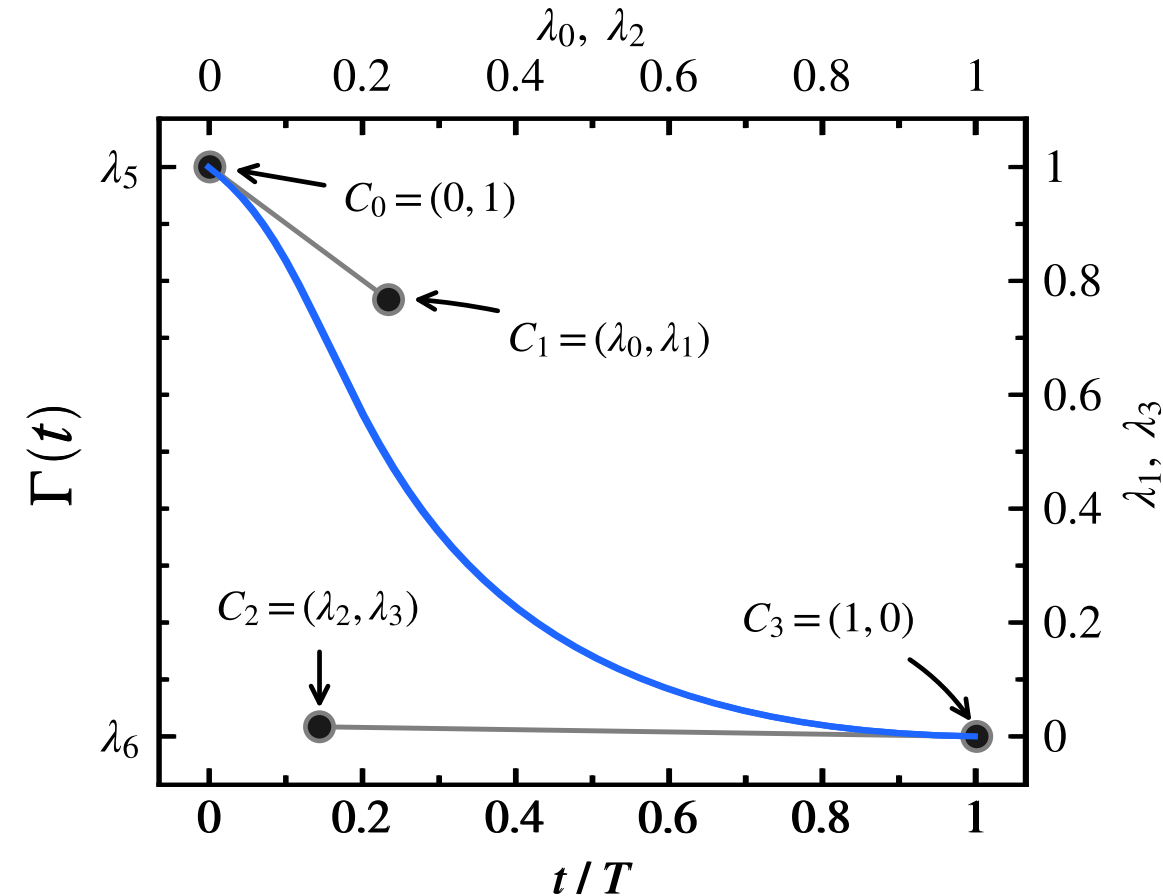
Bezier curves

➤ Optimize only 6 hyperparameters!



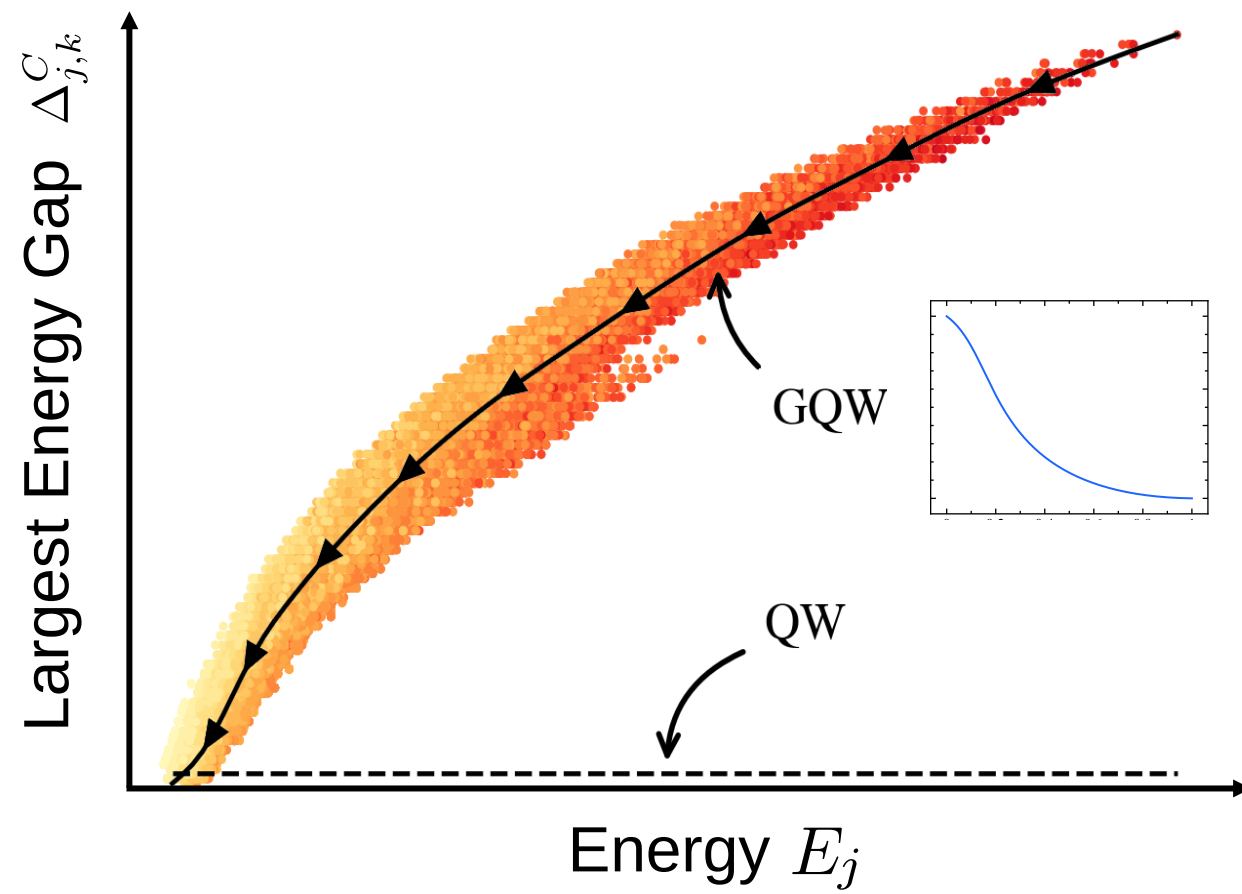
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LAT Theory: $\Gamma(t) \approx \langle \Delta^C \rangle(E(t)) / O(1)$

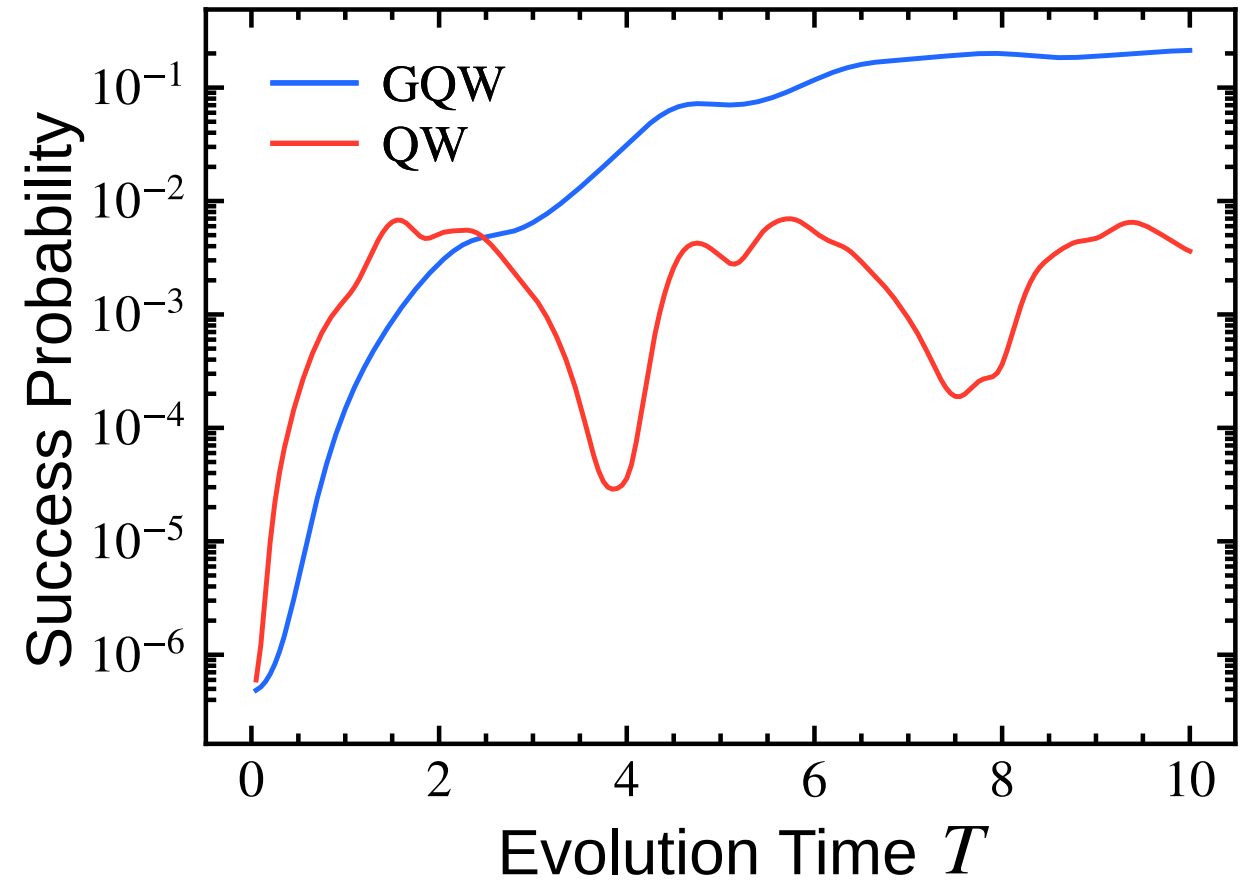


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Performance vs. Quantum Walk



$$H_{GQW}(t) = \Gamma(t) \cdot \left(-\sum_{i=1}^N \sigma_i^x \right) + \left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)$$

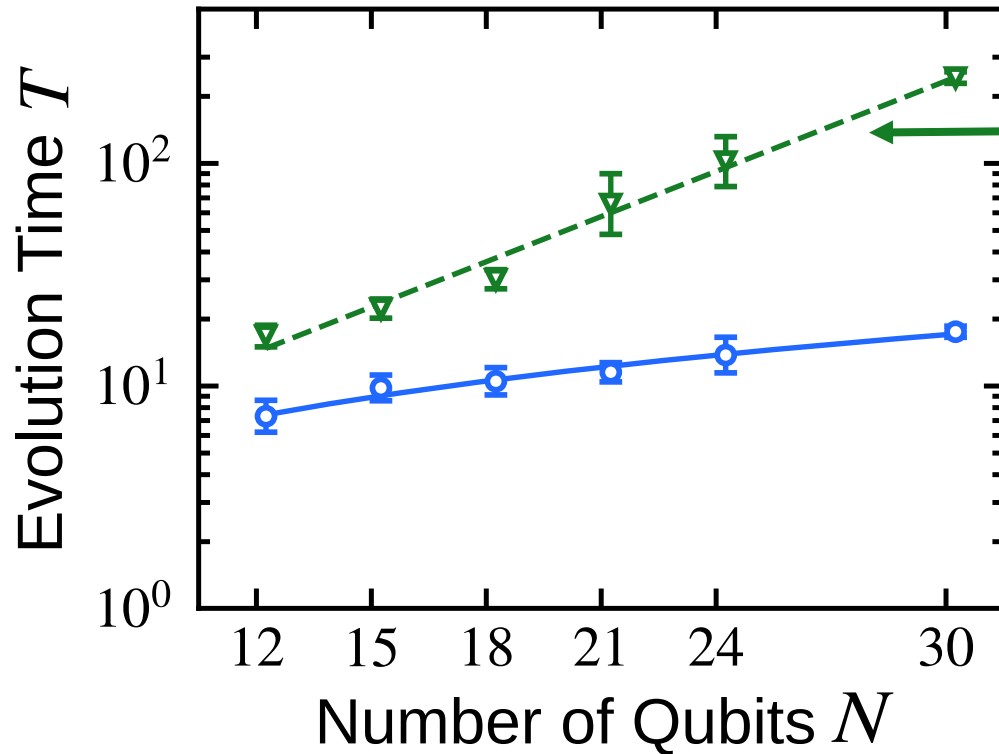


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Performance vs. Quantum Annealing

$$H_{GQW}(t) = \Gamma(t) \cdot \left(-\sum_{i=1}^N \sigma_i^x \right) + \left(-\sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)$$

Time to achieve 90% success probability:



QA: $T \sim 2^{0.22 N}$

GQW: $T \sim 0.54 N$

How can the scaling be **linear**?

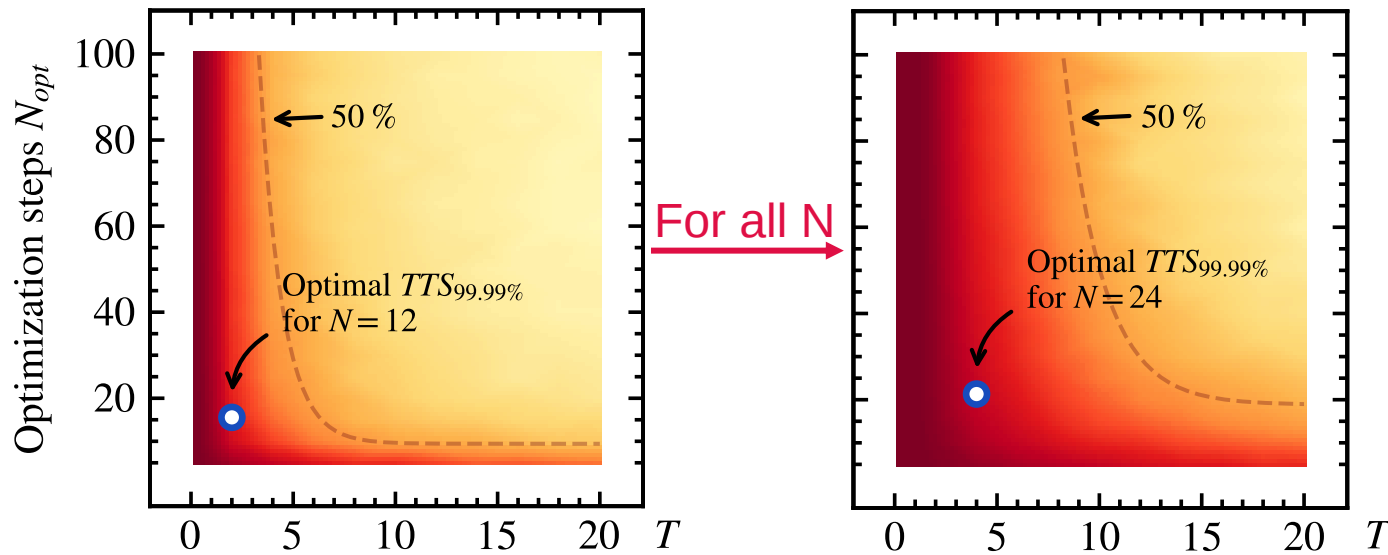
→ Need to properly account for the **cost of the classical optimization phase!**

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Accounting for Cost of Classical Optimizer

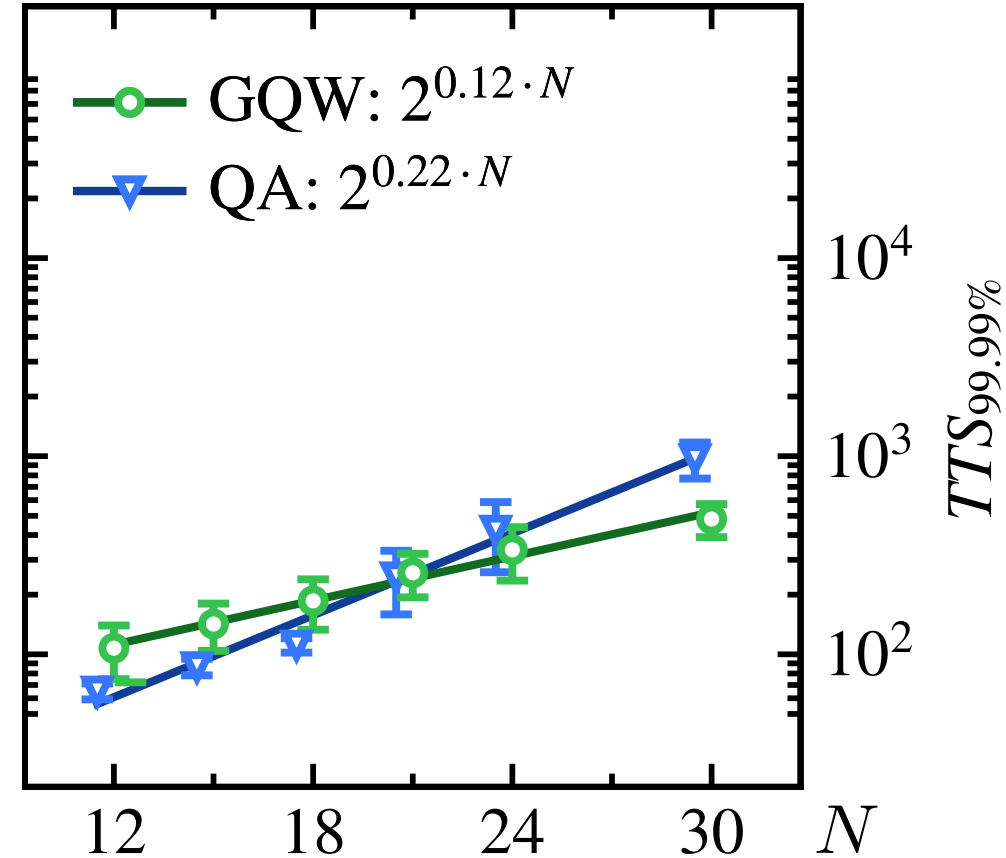
Time to achieve given success probability:

$$\text{TTS}_{P_{\text{target}}} = \frac{\ln(1 - P_{\text{target}})}{\ln(1 - P_{\text{groundstate}})} T + N_{\text{opt}} T$$



Proper accounting recovers **exponential scaling**

$$H_{GQW}(t) = \Gamma(t) \cdot \left(- \sum_{i=1}^N \sigma_i^x \right) + \left(- \sum_{i=1}^N h_i \sigma_i^z + \sum_{i \neq j} J_{ij} \sigma_i^z \sigma_j^z \right)$$



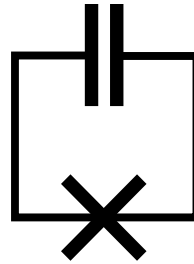
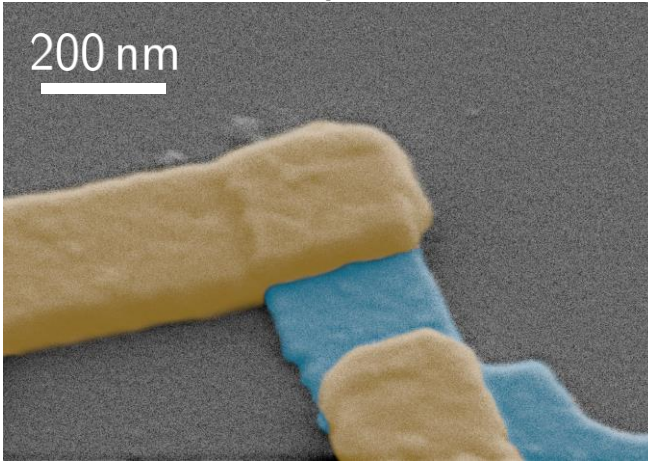
→ **GQW scales favourably without cheating**

INTERLUDE:

JOSEPHSON HARMONICS

Using the Right Model to Describe Data

Standard Josephson Junction



$$H = 4E_c n^2 - \cancel{E_J \cos \varphi} - \sum_m E_{Jm} \cos m\varphi$$

Paper: arXiv:2302.09192

Talk: <https://youtu.be/dWu3lCwyXBM>

Observation of Josephson Harmonics in Tunnel Junctions

Dennis Willsch,^{1, *} Dennis Rieger,^{2, 3, *} Patrick Winkel,^{2, 3, 4, 5} Madita Willsch,^{1, 6} Christian Dickel,⁷ Jonas Krause,⁷ Yoichi Ando,⁷ Raphaël Lescanne,^{8, 9} Zaki Leghtas,⁸ Nicholas T. Bronn,¹⁰ Pratiti Deb,¹⁰ Olivia Lanes,¹⁰ Zlatko K. Minev,¹⁰ Benedikt Dennig,^{2, 3} Simon Geisert,² Simon Günzler,² Sören Ihssen,² Patrick Paluch,^{2, 3} Thomas Reisinger,² Roudy Hanna,^{11, 12} Jin Hee Bae,¹¹ Peter Schüffelgen,¹¹ Detlev Grützmacher,^{11, 12} Luiza Buimaga-Iarinca,¹³ Cristian Morari,¹³ Wolfgang Wernsdorfer,^{2, 3} David P. DiVincenzo,^{14, 12} Kristel Michielsen,^{1, 6, 12} Gianluigi Catelani,^{15, 16} and Ioan M. Pop^{2, 3, †}

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Generalization of Schedules to Larger Problems

- Optimize GQW schedule $\Gamma(t)$ for $N=30$
- Apply same schedule until $N=40$ qubits
- Comparison vs.

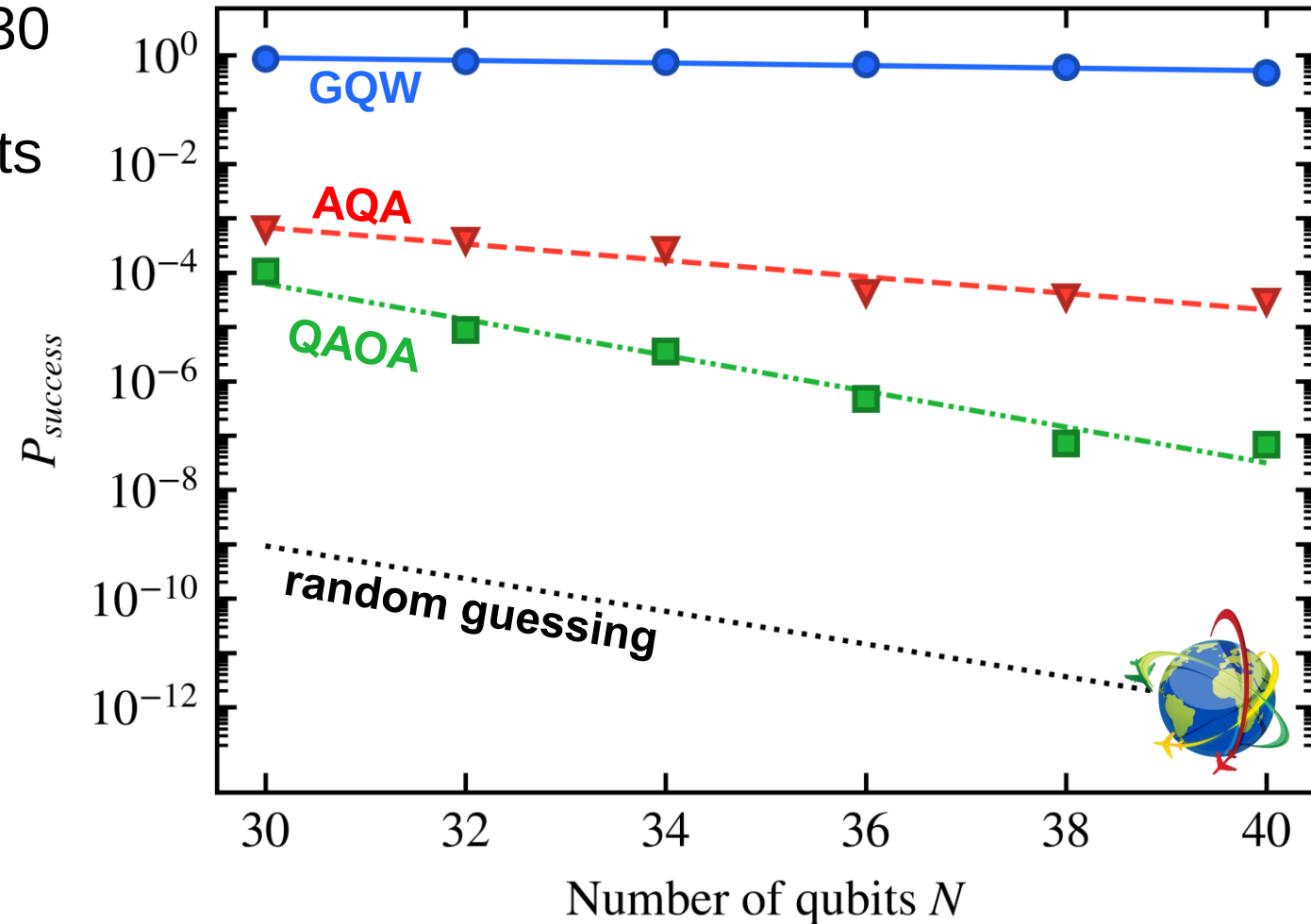
QAOA (2p = 12 parameters)

$$|\beta, \gamma\rangle = e^{-i\beta_p H_D} e^{-i\gamma_p H_C} \dots e^{-i\beta_1 H_D} e^{-i\gamma_1 H_C} |+\rangle^{\otimes N}$$

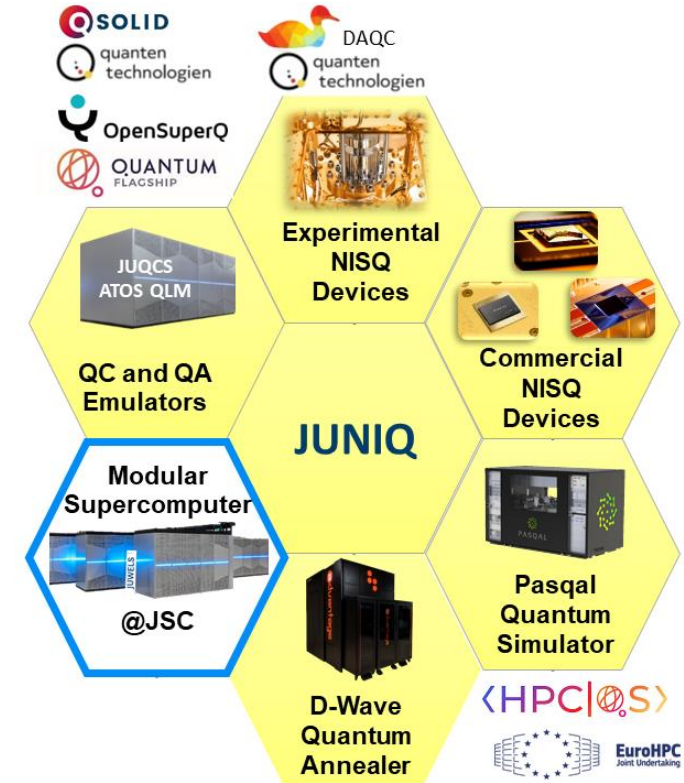
AQA (1 parameter) [arXiv:2104.03293](https://arxiv.org/abs/2104.03293)

$$\begin{aligned} \beta_k &= -\tau(A(s_{k+1}) + A(s_k))/2, & k &= 1, \dots, p-1, \\ \beta_p &= -\tau A(s_p)/2, \\ \gamma_k &= \tau B(s_k), & k &= 1, \dots, p, \end{aligned}$$

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THANK YOU FOR YOUR ATTENTION



JUNIQ Rolling Call:



or search for
“FZJ JUNIQ ACCESS”