



UNIVERSITÀ
DI TRENTO



Quantum Science and Technology in Trento



Trento Institute for
Fundamental Physics
and Applications

TRAPPED-ION QUANTUM COMPUTING FOR COLLECTIVE NEUTRINO OSCILLATIONS

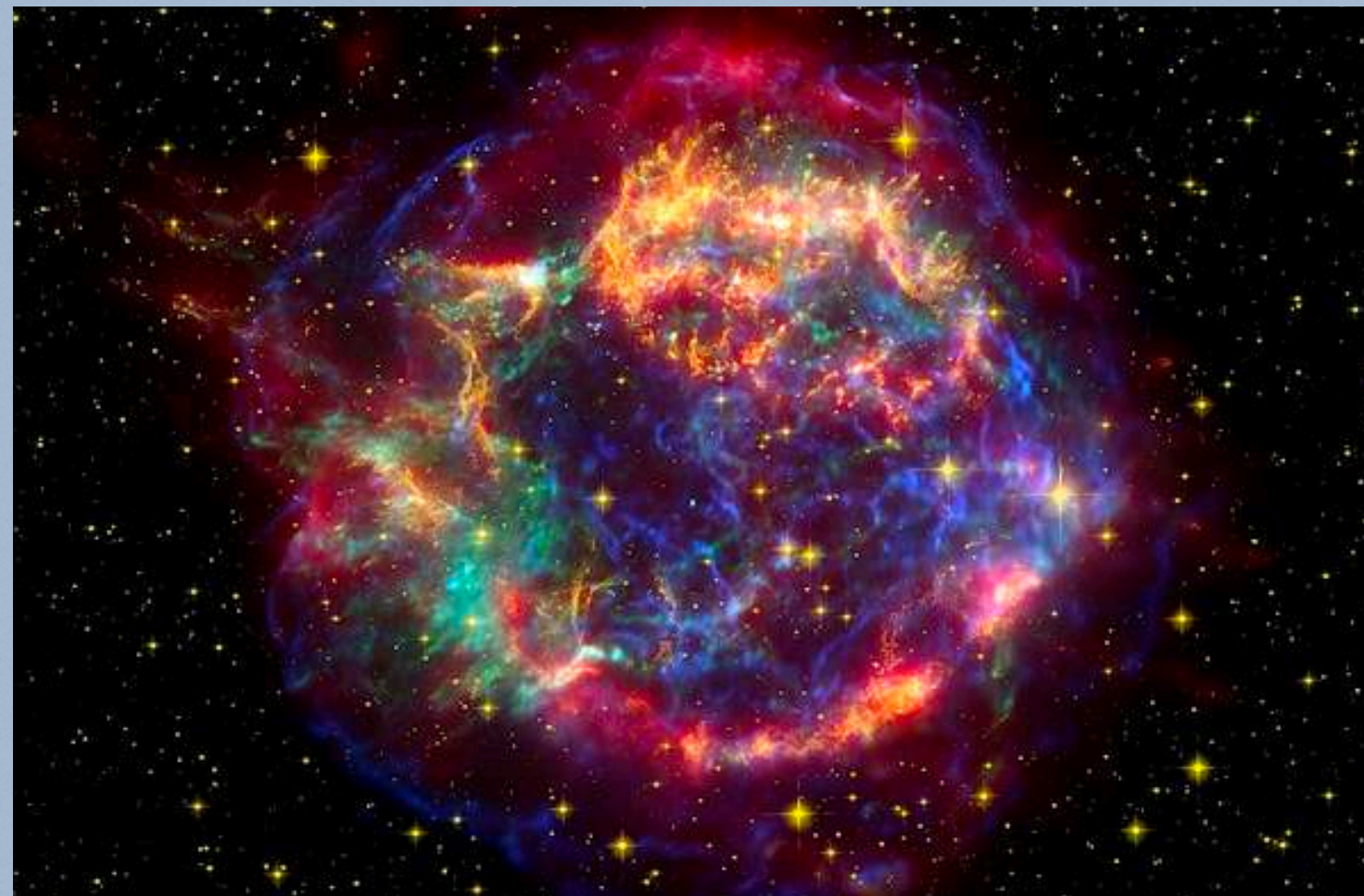
Valentina Amitrano
Francesco Pederiva
Alessandro Roggero

Nuclear and Particle Physics on a Quantum Computer:
Where do we stand now?

OUTLINE

Introduction

- Motivation
- Physical description of the many-neutrino system in high density environment



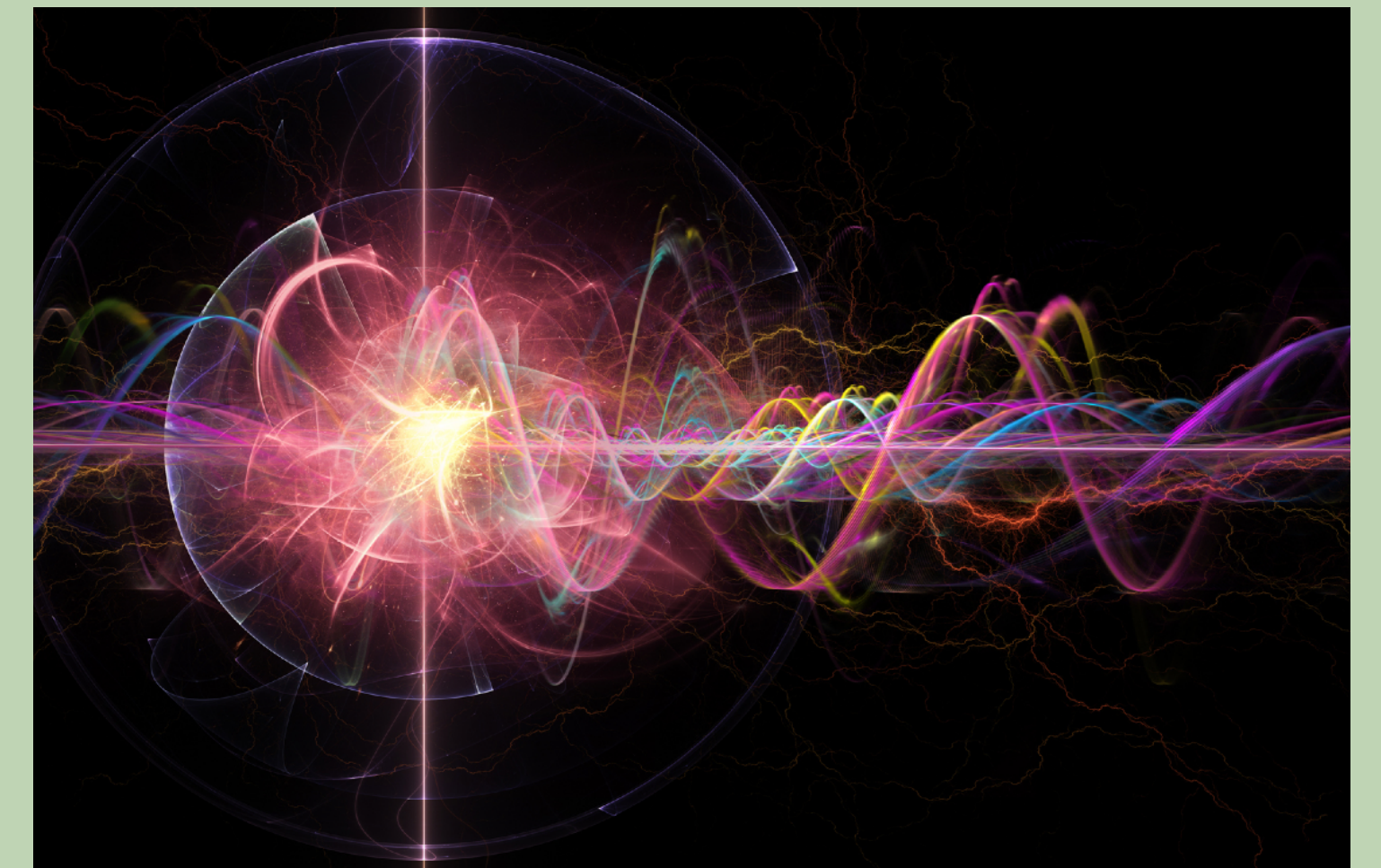
QC simulation

- Hamiltonian simulation
- The quantum algorithm implementation

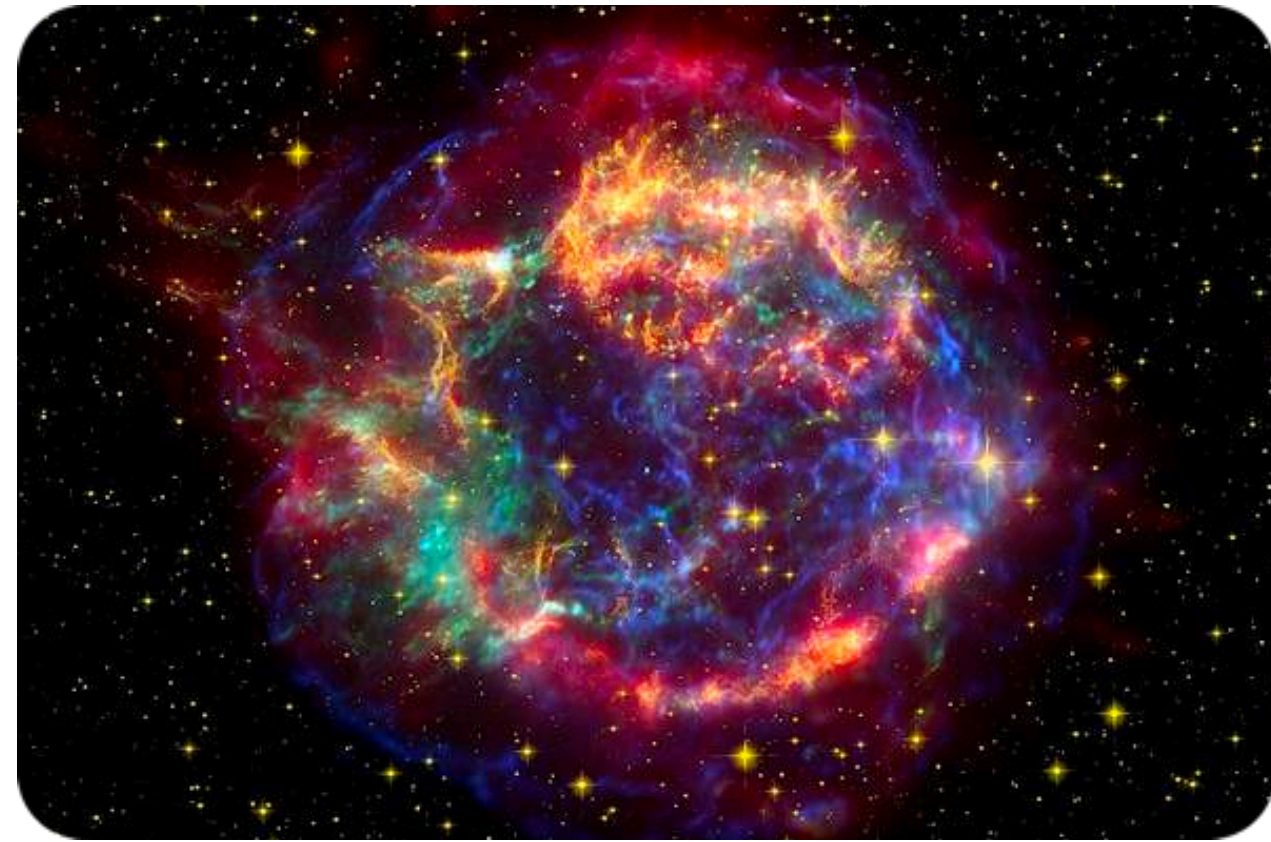


Results

- Data from the real trapped-ion quantum machine:
Quantinuum System Model

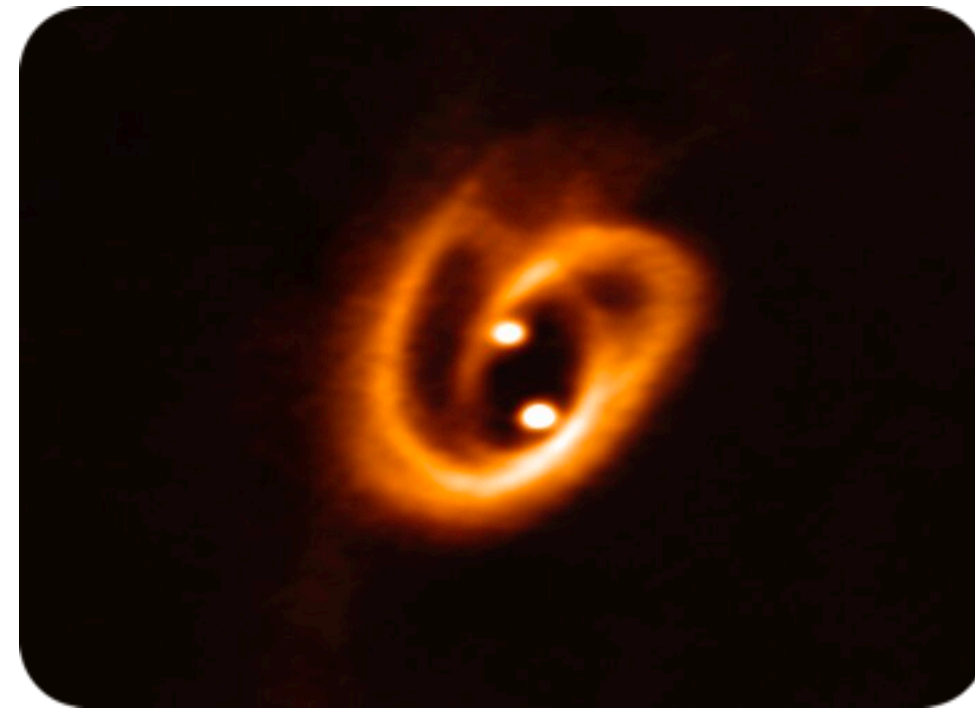


WHY WE CARE ABOUT NEUTRINOS

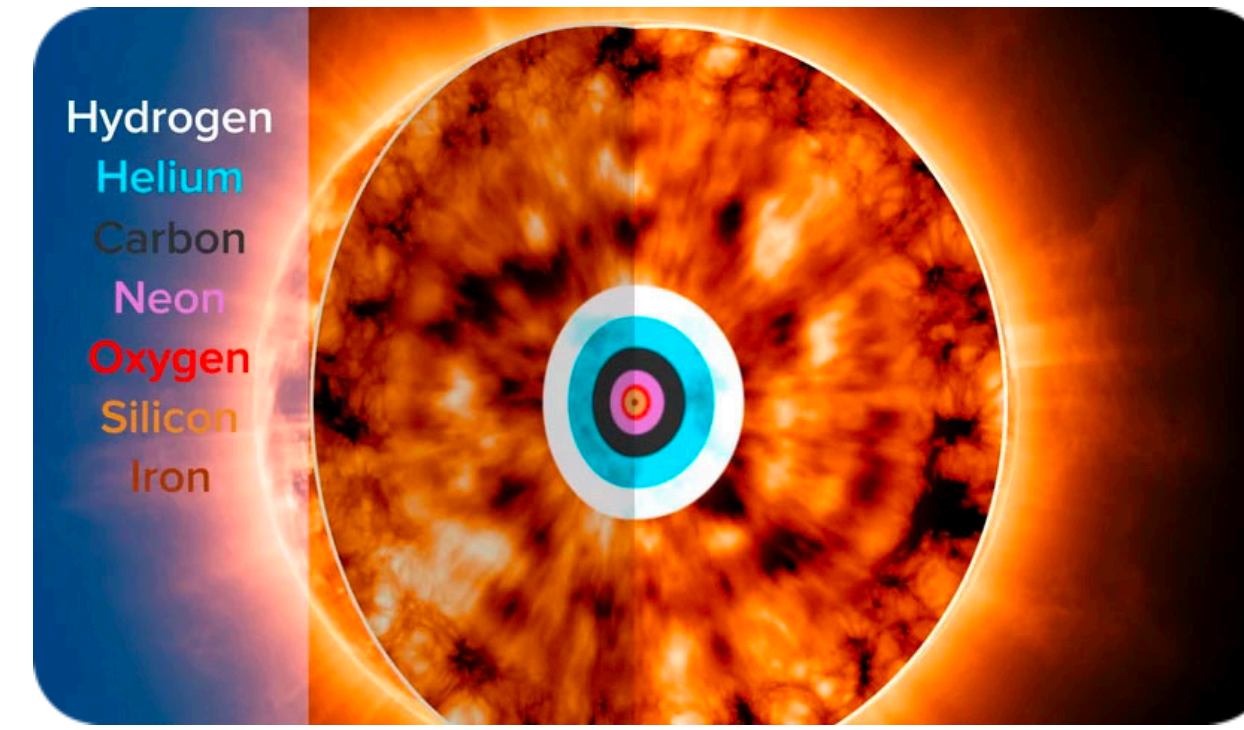


Core-collapse supernovae

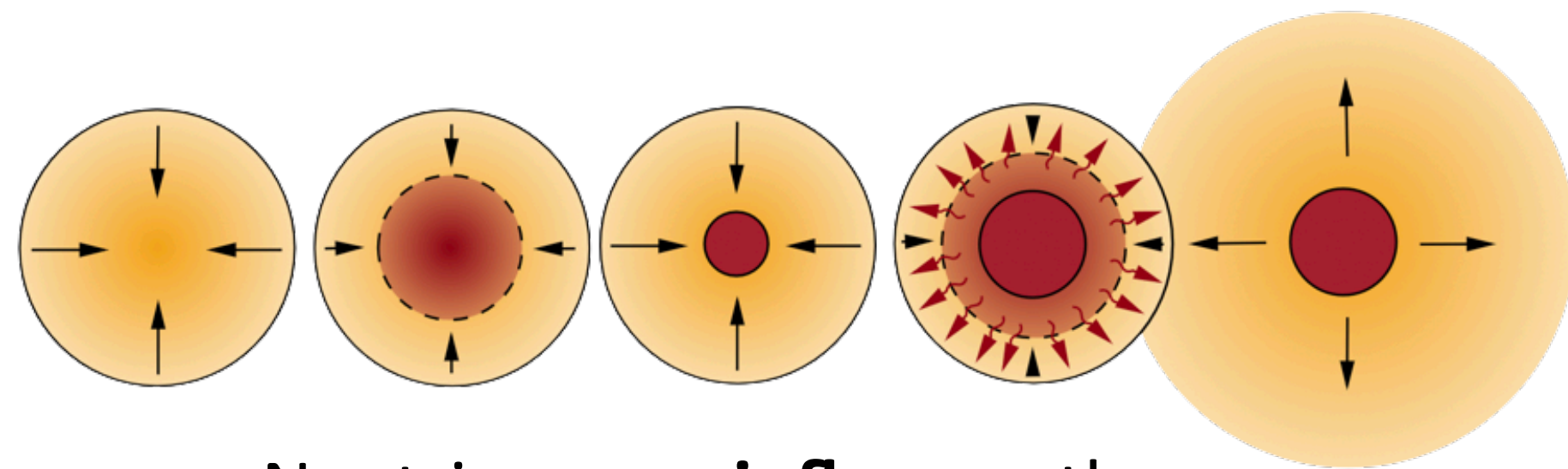
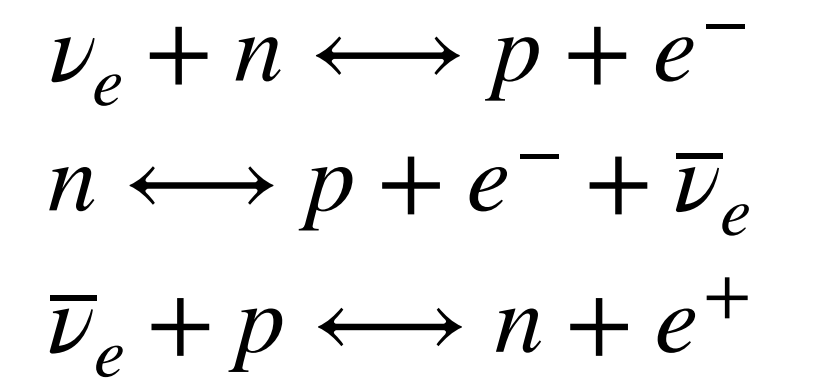
Neutrinos are **messengers of information** of physics under extreme conditions



Massive star mergers



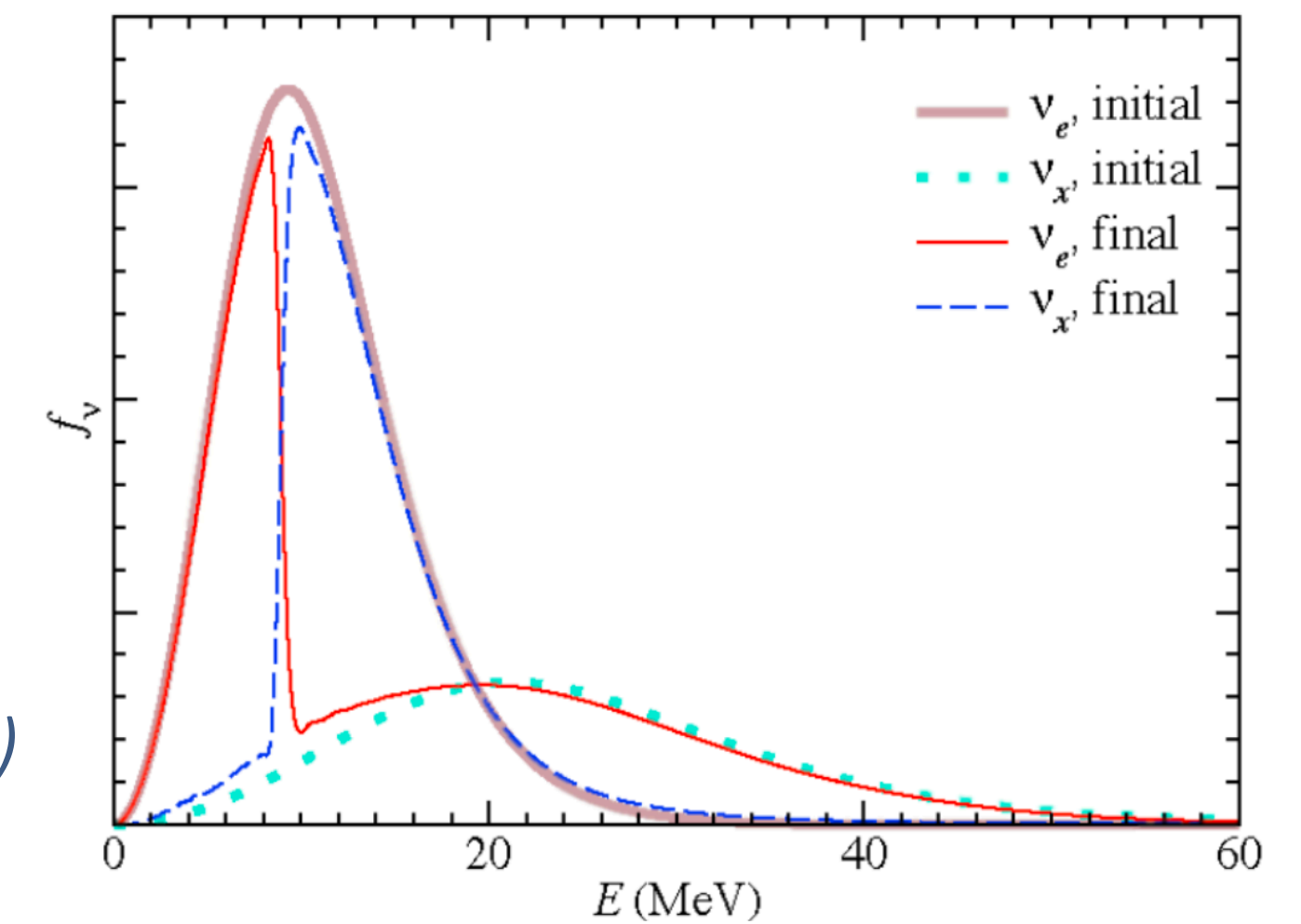
Nucleosynthesis and in general weak interaction is **flavor-dependent**



Neutrinos can **influence** the supernovae **explosion**

Spectral splits can happen at some distance from the emission sphere

Duan et. al (2006)



NEUTRINOS FROM CORE-COLLAPSE SUPERNOVAE

- Massive stars $M \geq 8 M_{\odot}$ explode releasing a huge amount of energy and neutrinos $\sim 10^{58}$
- Flavor Hamiltonian of many-neutrino system

$$H = H_{vac} + H_{\nu e} + H_{\nu\nu}$$

Vacuum:

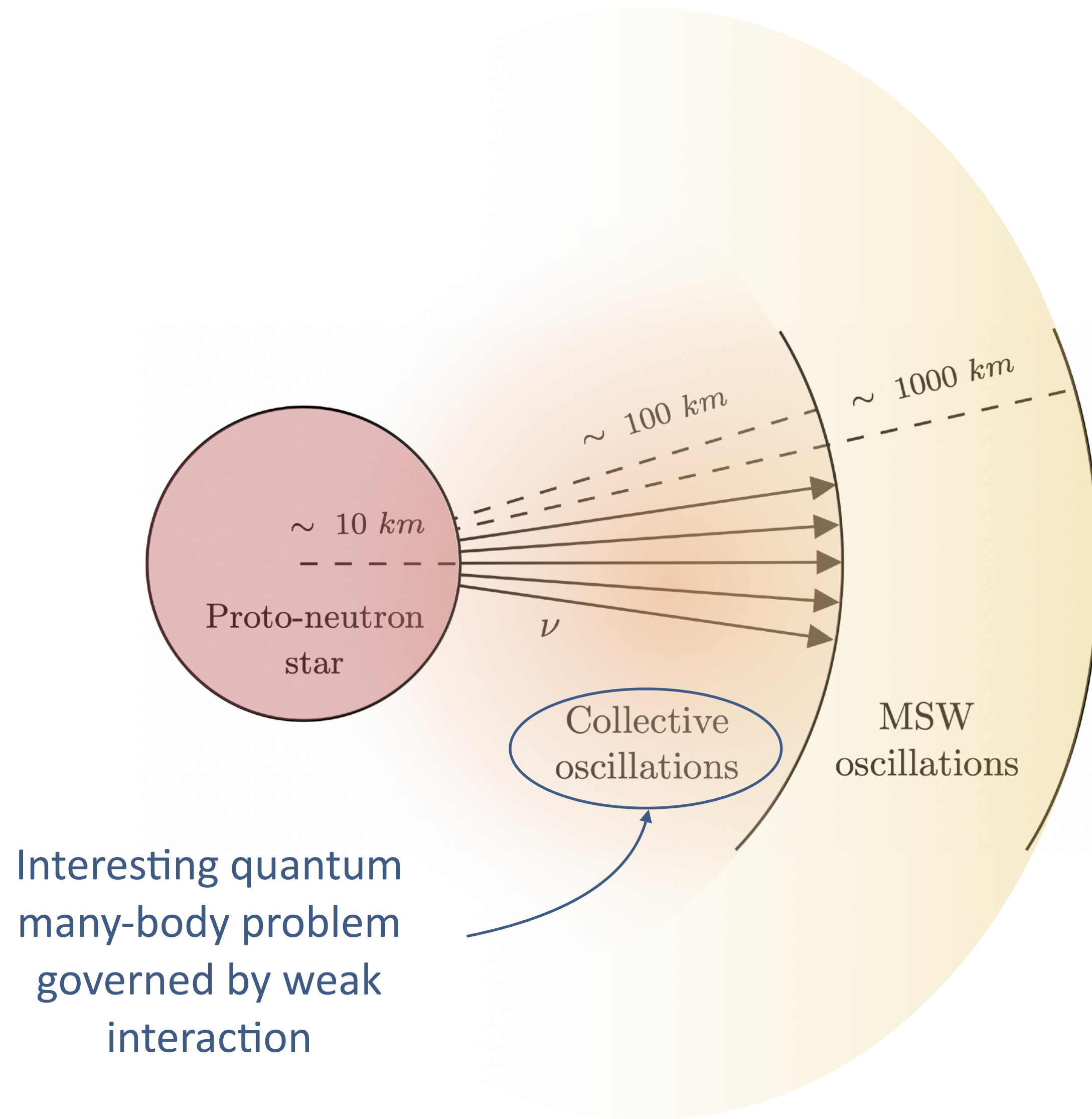
Mass eigenstates \neq
flavor eigenstates

MSW:

Scattering
with matter

$\nu\nu$ -interaction:

Forward
scattering



NEUTRINOS FROM CORE-COLLAPSE SUPERNOVAE

Two-flavor Hamiltonian (SU(2) model): the flavor state of a neutrino is a **flavor isospin** $|\nu\rangle = \alpha|\nu_e\rangle + \beta|\nu_x\rangle$

$$H = H_{vac} + H_{\nu\nu}$$

1-body term

2-body term

$$H_{vac} = \Delta \sum_{i=1}^N \vec{b} \cdot \vec{\sigma}_i$$

$$\Delta = \frac{\delta m^2}{4E}$$

$$\vec{b} = (\sin(2\theta_\nu), 0, -\cos(2\theta_\nu))$$

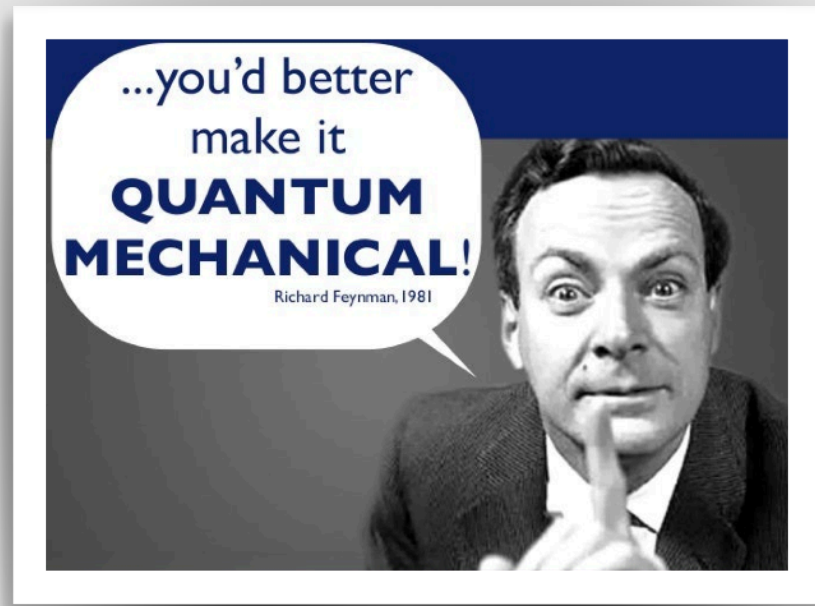
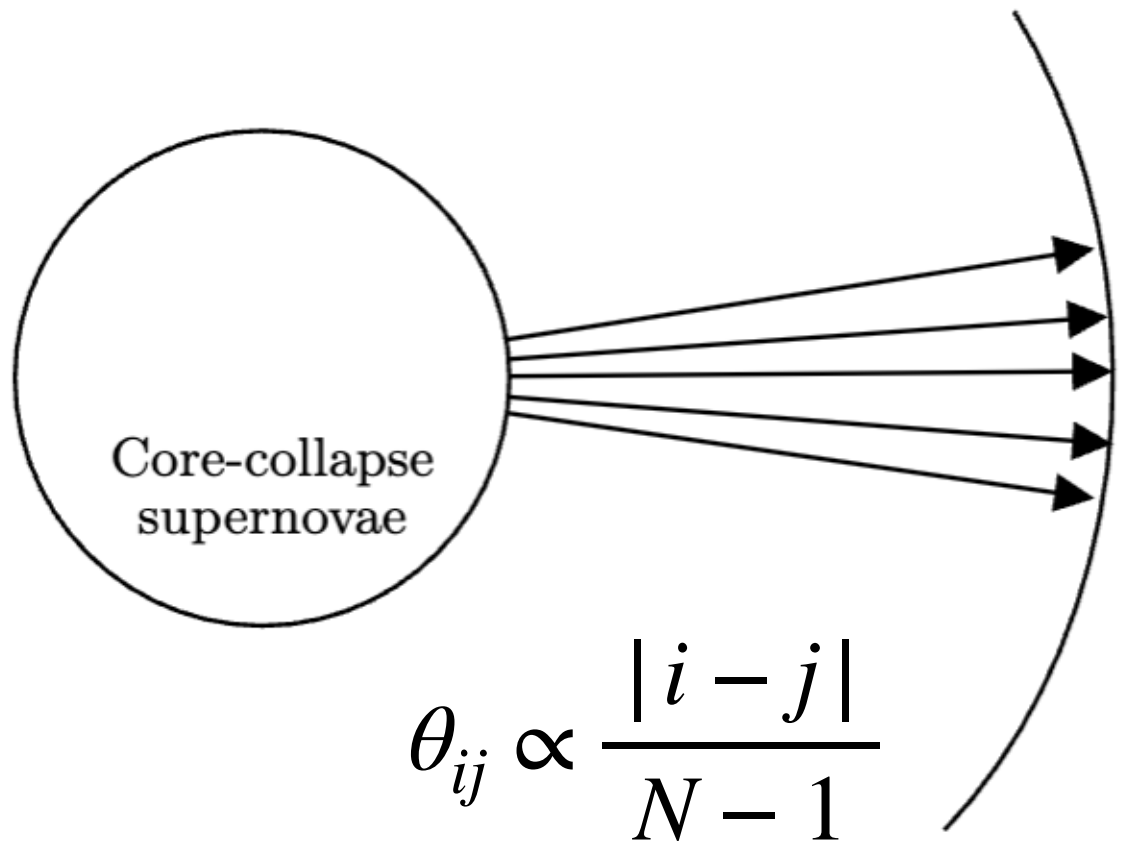
$$\vec{\sigma} = (X, Y, Z)$$

$$H_{\nu\nu} = \frac{\mu}{N} \sum_{i<j}^N J_{ij} \vec{\sigma}_i \cdot \vec{\sigma}_j$$

$$\mu = \sqrt{2} G_F n_\nu$$

$$J_{ij} = 1 - \cos(\theta_{ij})$$

$$\cos(\theta_{ij}) = \frac{\vec{p}_i \cdot \vec{p}_j}{\|\vec{p}_i\| \|\vec{p}_j\|}$$



Simulating the full dynamics is difficult using classical resource

THE THEORETICAL EVOLUTION

We want to simulate the flavor evolution

- Initial state $|\Psi_0\rangle = |\nu_e\rangle^{\otimes N/2} \otimes |\nu_x\rangle^{\otimes N/2}$
- Evolved state $|\Psi(t)\rangle = e^{-iHt} |\Psi_0\rangle$
- $\langle \nu_e | Z | \nu_e \rangle = 1$ and $\langle \nu_x | Z | \nu_x \rangle = -1$
- Measure the probability to be in the inverted flavor as a function of time

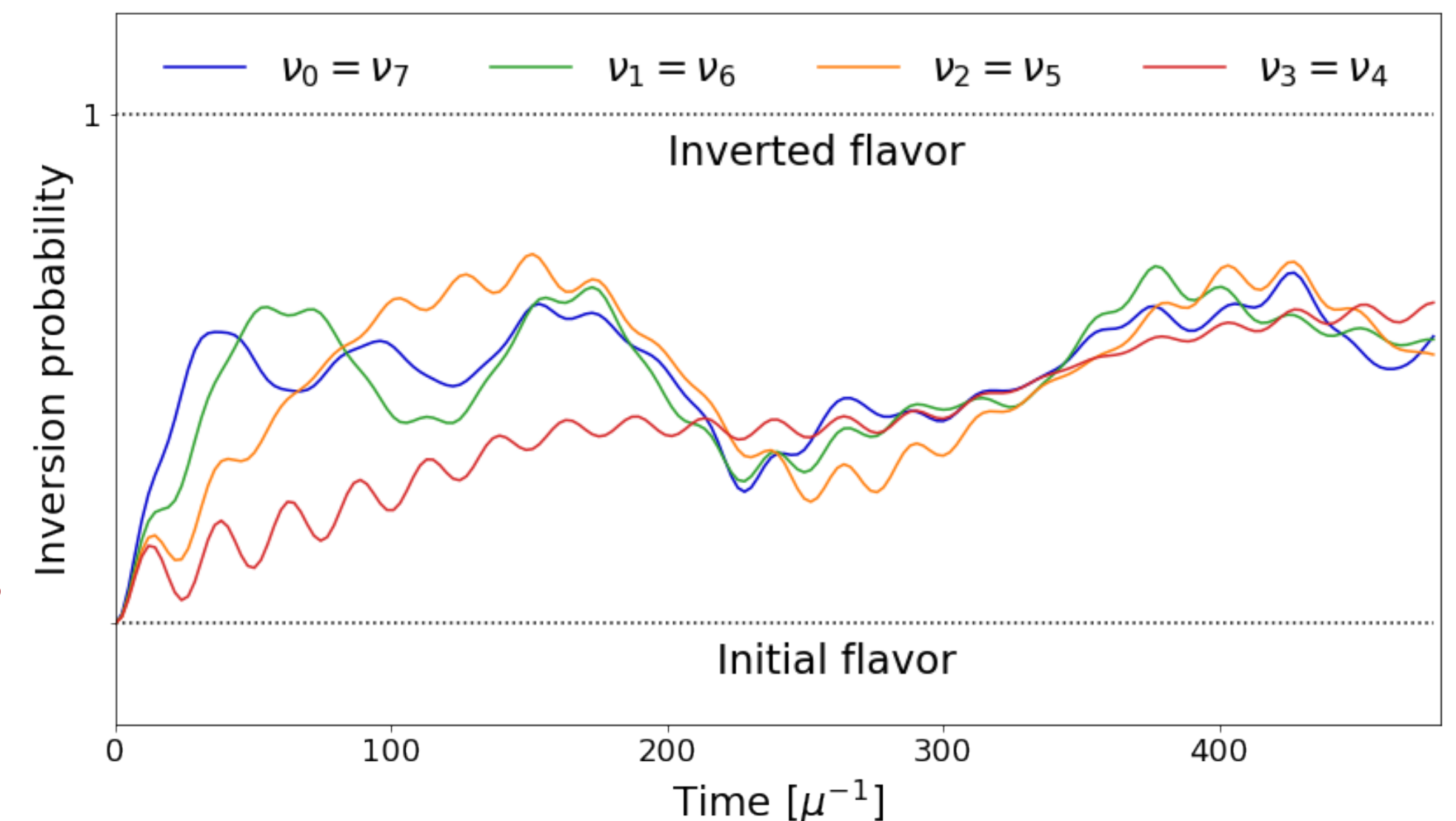
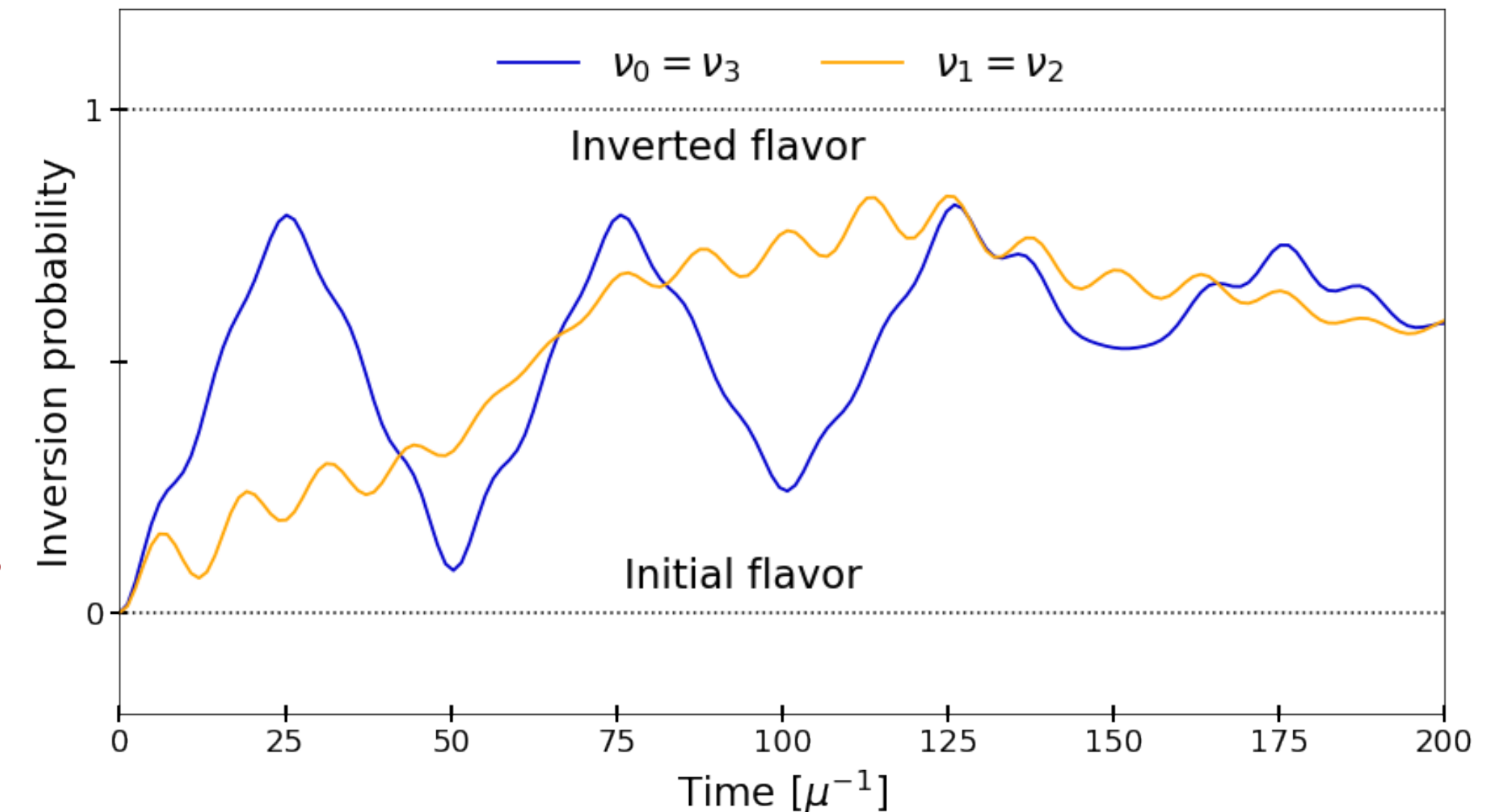
$$P_{inv}^{(i)}(t) = \frac{|\langle Z_i(0) \rangle - \langle Z_i(t) \rangle|}{2}$$

- Note the symmetry under particle exchange
 - Symmetric Hamiltonian
 - Anti-symmetric initial state
 - $\nu_k \longleftrightarrow \nu_{N-1-k}$

$N = 4$
 $2^4 \times 2^4$ matrix
 multiplications

$N = 8$
 $2^8 \times 2^8$ matrix
 multiplications

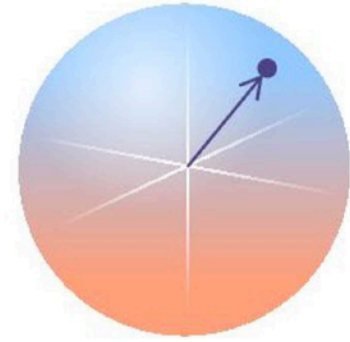
V. Amitrano et. al. Phys. Rev. D 107, 023007 (2023)



INGREDIENTS FOR HAMILTONIAN SIMULATION

1° ingredient:
Encoding map

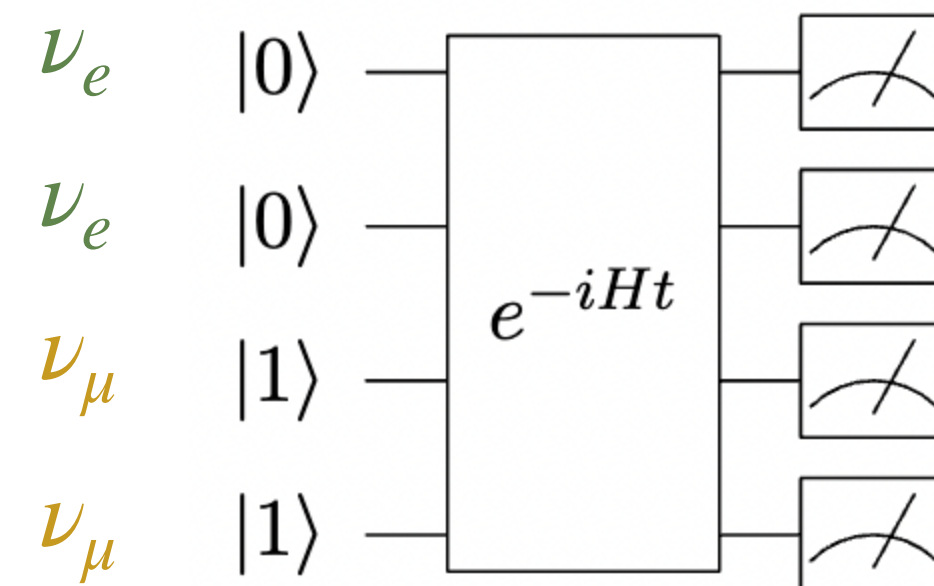
$$|\nu_e\rangle \mapsto |0\rangle$$



$$|\nu_x\rangle \mapsto |1\rangle$$

- Two-flavor approximation $|\nu\rangle = \alpha|\nu_e\rangle + \beta|\nu_x\rangle$
- Qubit state $|\nu\rangle = \alpha|0\rangle + \beta|1\rangle$
- N neutrinos encoded into N qubits

2° ingredient:
Unitary implementation



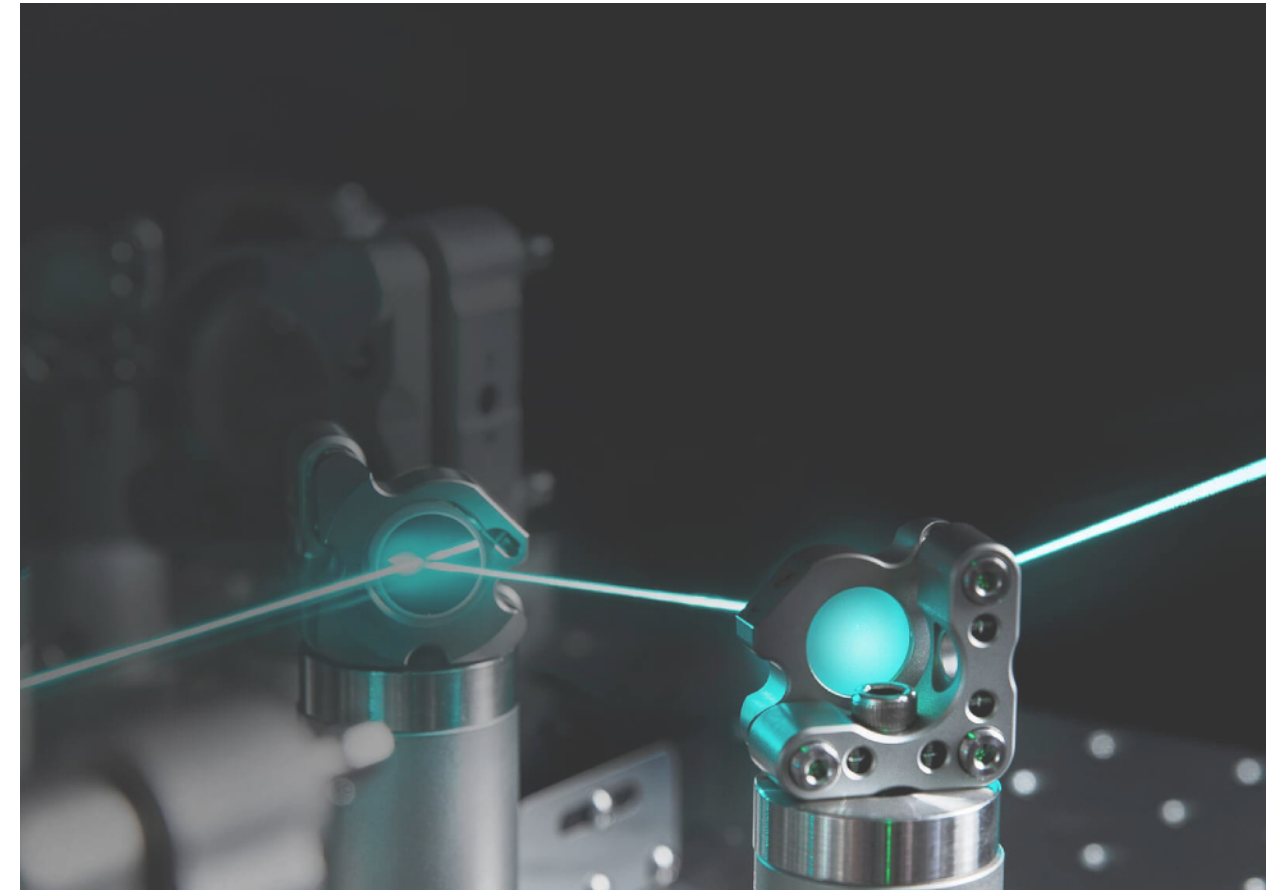
- Implement the propagator $U(t) = e^{-iHt}$ generated by the Hamiltonian

$$H = \sum_i \vec{b} \cdot \vec{\sigma}_i + \sum_{i<j} J_{ij} \vec{\sigma}_i \cdot \vec{\sigma}_j$$

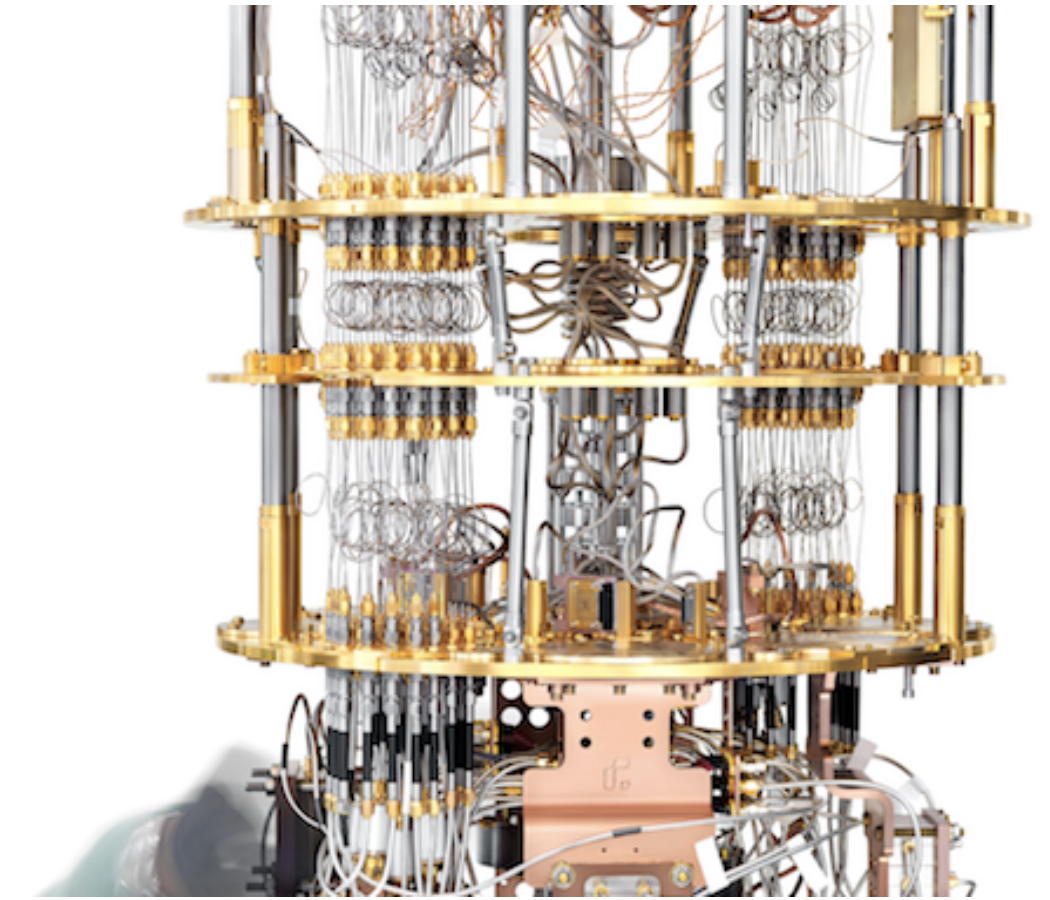
- **Quantum gate decomposition** procedure to obtain a quantum circuit
- Exponential number of operations in general... we need to optimize it!
- All-to-all interactions are difficult with reduced connectivity

THE UNITARY IMPLEMENTATION: MACHINE AWARE COMPILATION

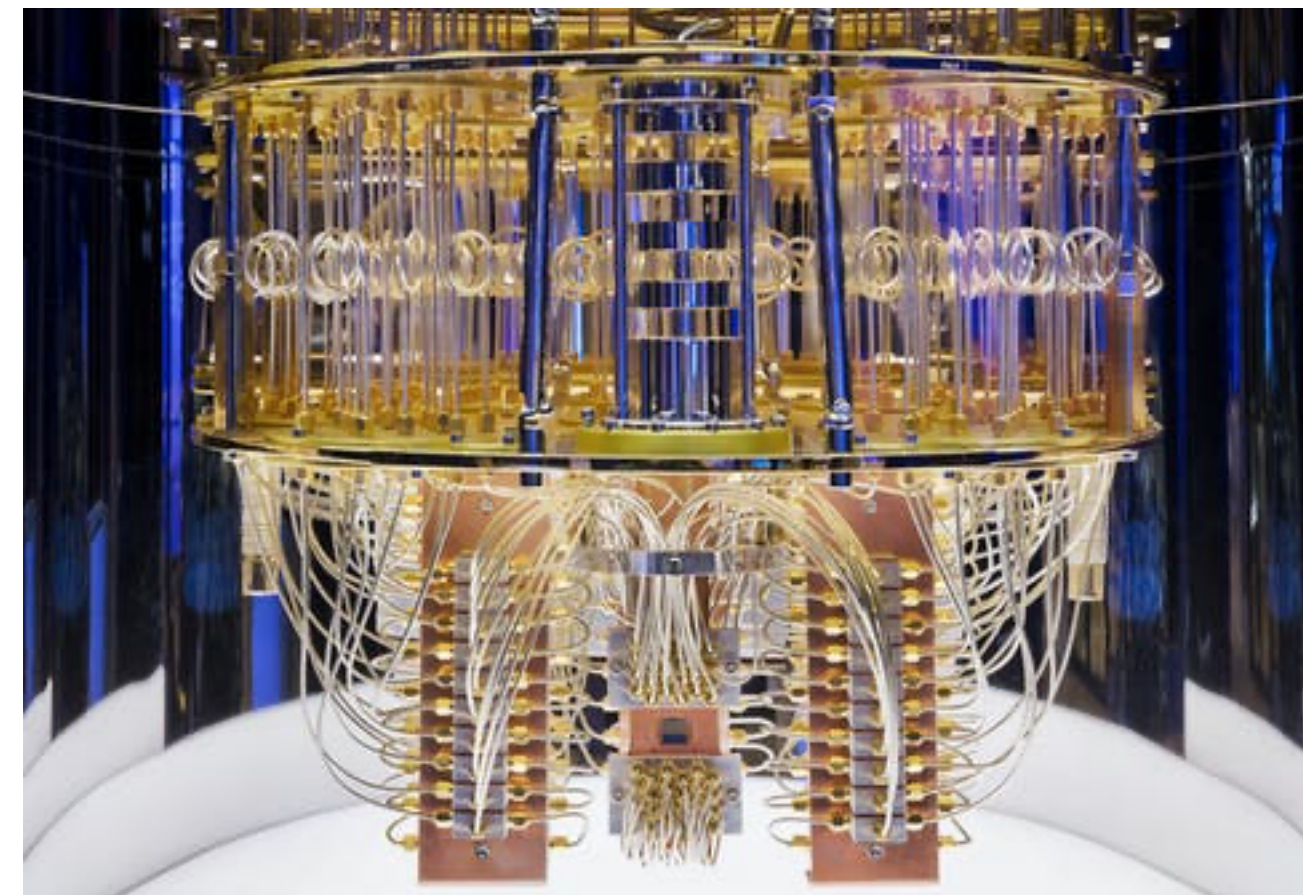
- Different **qubit**
 - Superconductive circuit
 - Trapped ions
- Different universal **gate set**
 - Circuit optimization
 - More control on what we are running
- Different qubit **connectivity**
 - Linear
 - All - to - all
 - Etc...



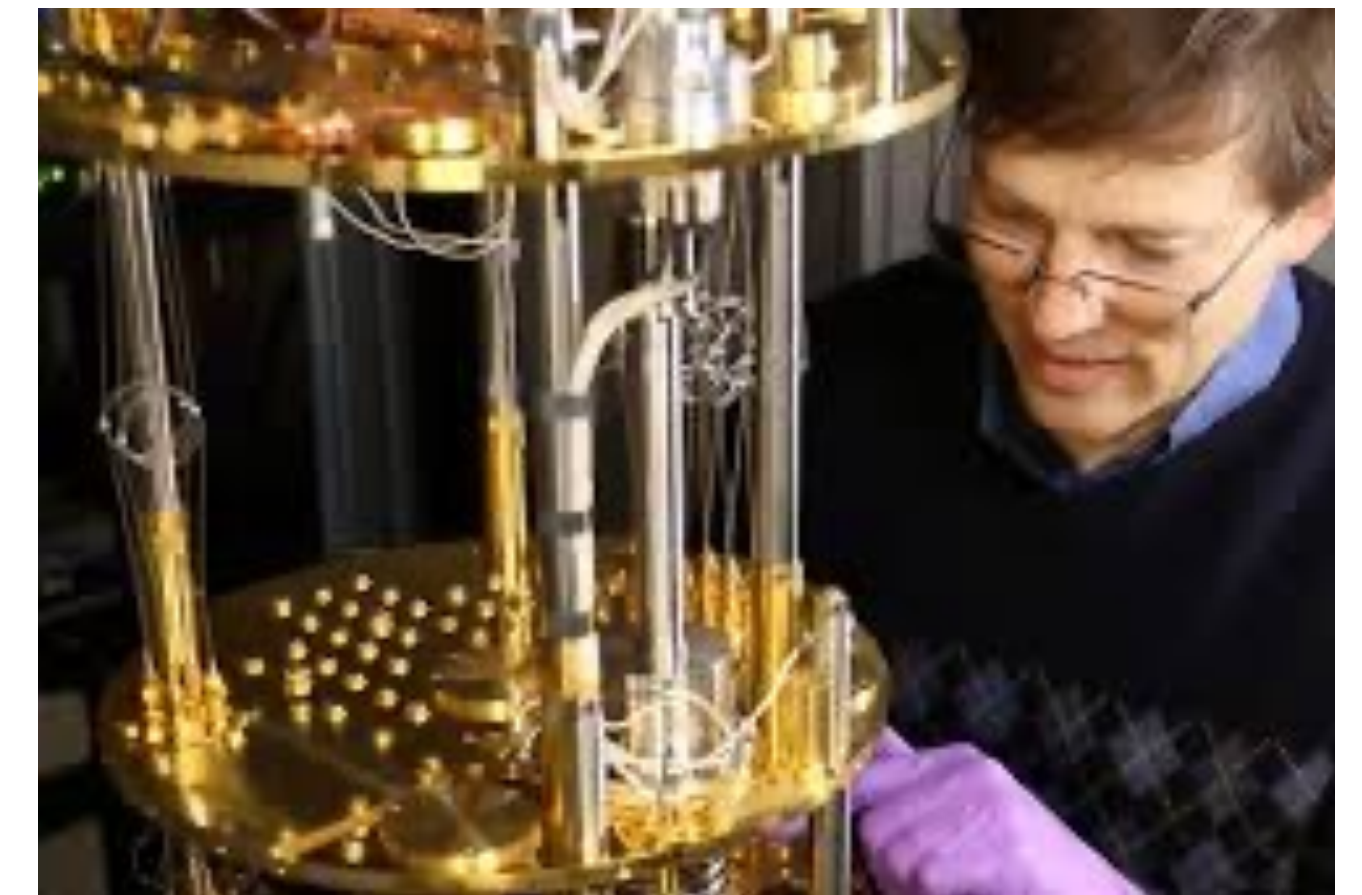
Honeywell Quantum



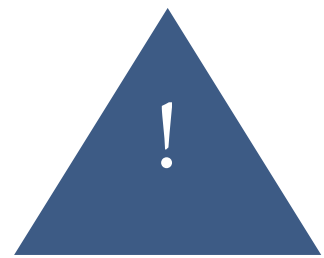
Rigetti Quantum



IBM Quantum

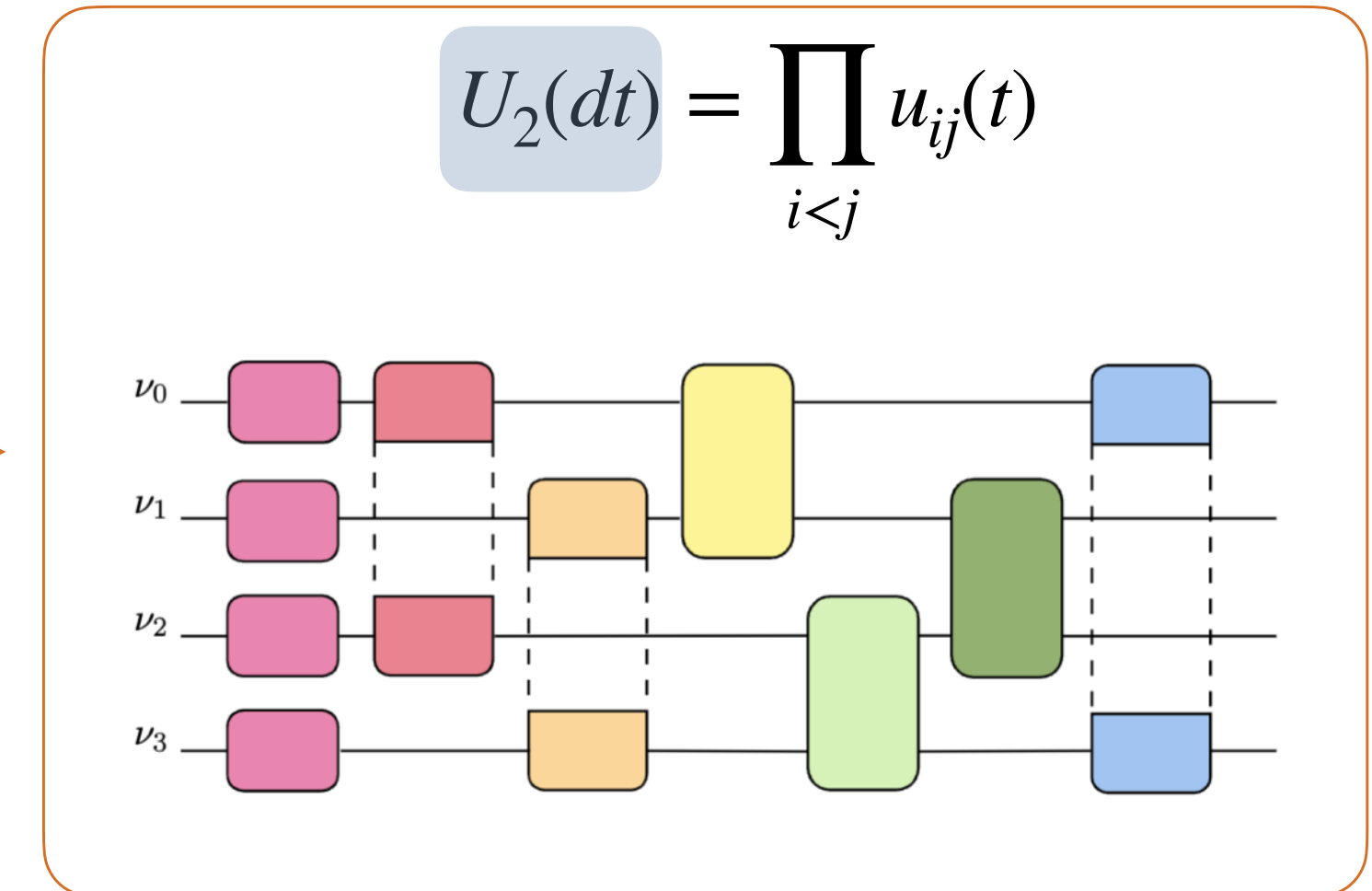
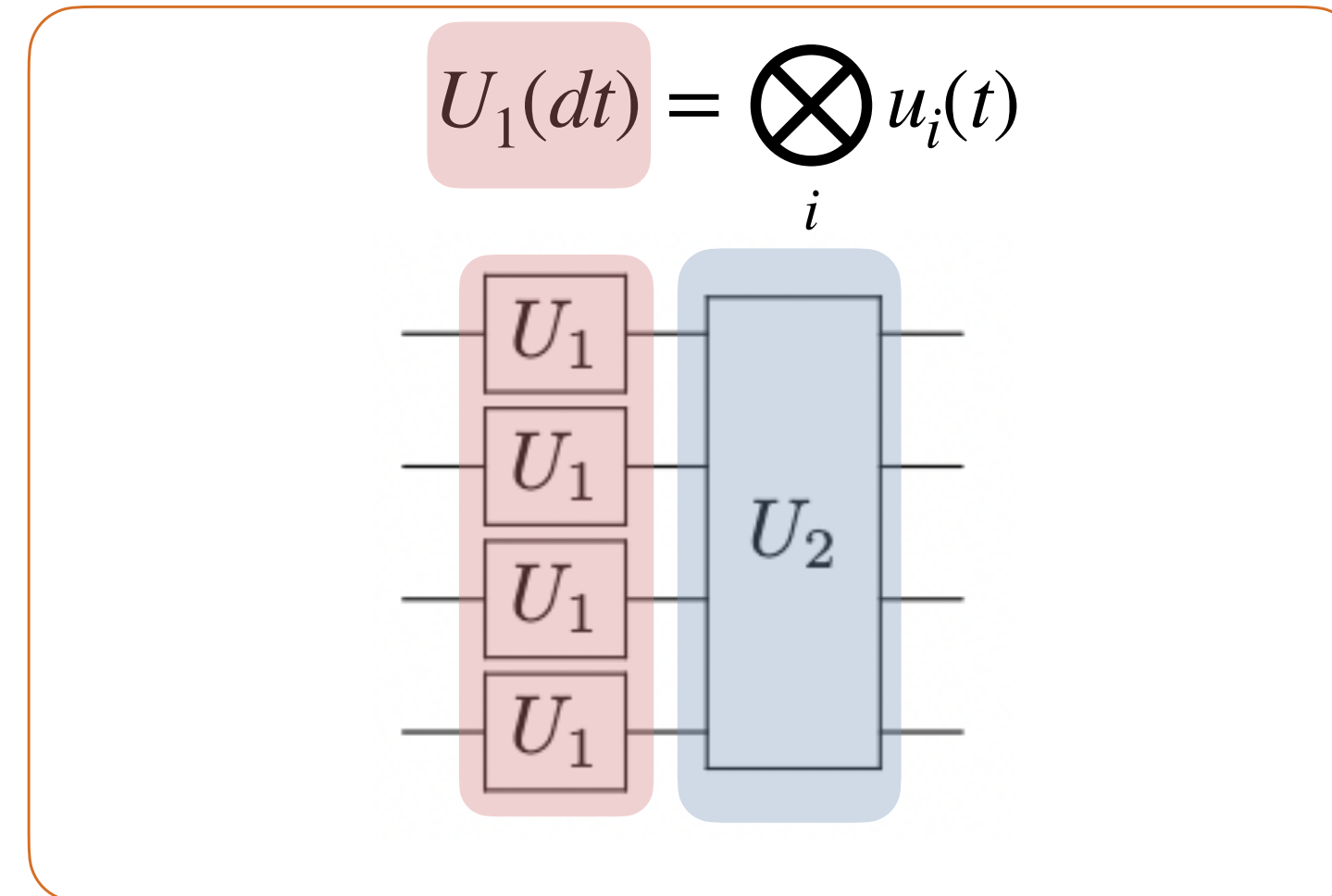
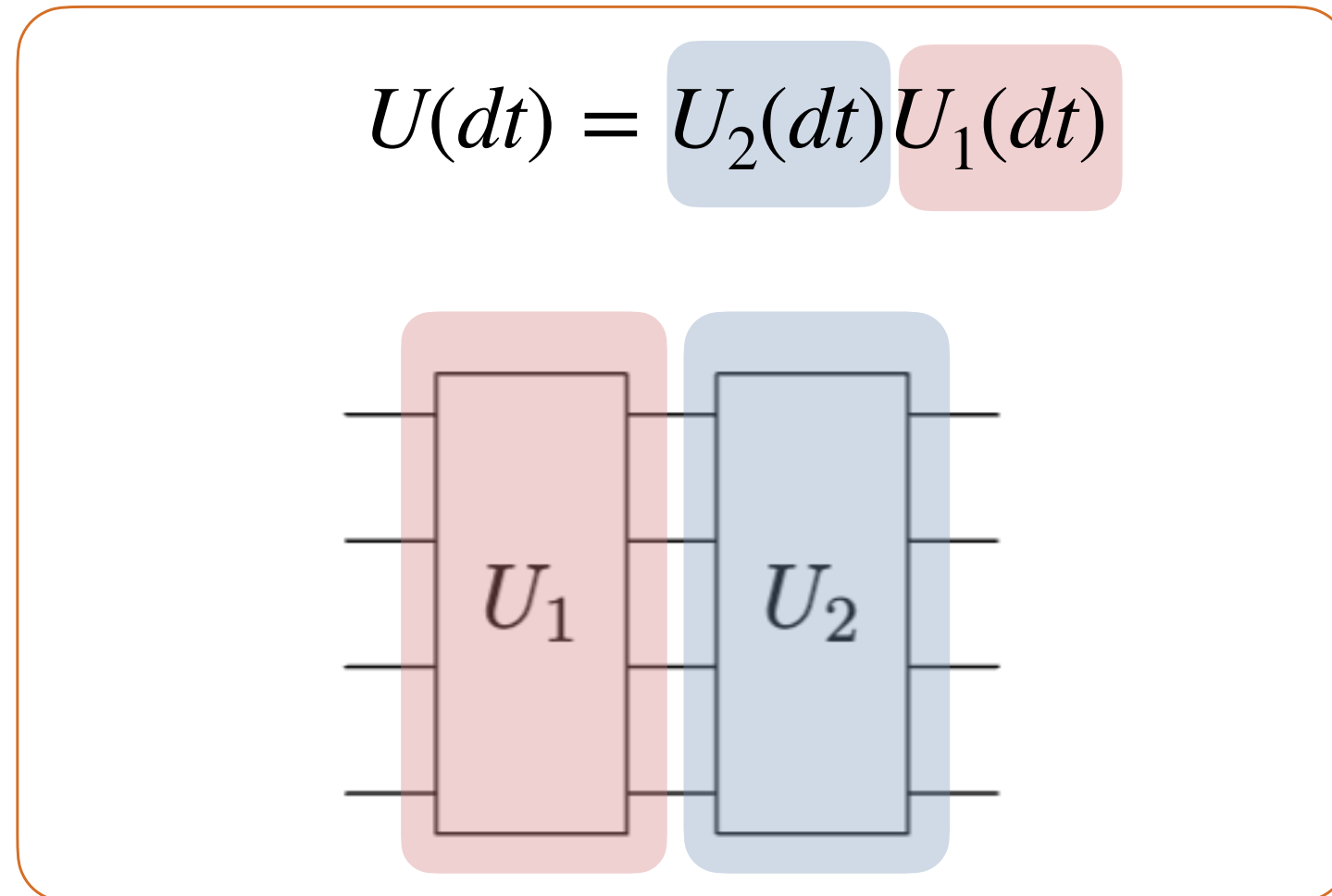


LLNL testbed



Trapped ions are perfect for the collective neutrino problem

THE UNITARY IMPLEMENTATION: GATE DECOMPOSITION



The total hamiltonian is

$U(dt) = e^{-i(H_{vac} + H_{\nu\nu})dt}$ and we can split 1-body and 2-body parts without error because $[H_{vac}, H_{\nu\nu}] = 0$

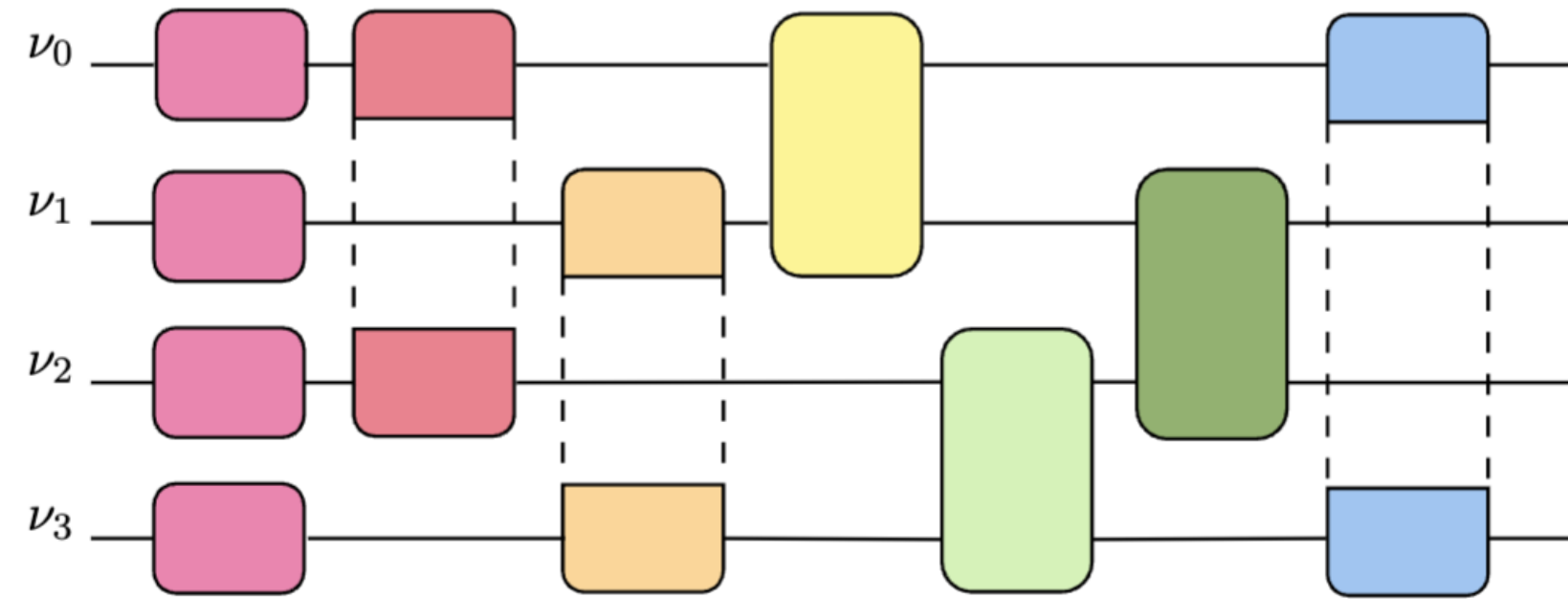
The 1-body part is simple

$U_1(dt) = e^{-iH_{vac}dt} = e^{-i\sum_i h_i dt}$
 where $h_i = \vec{b} \cdot \vec{\sigma}_i$ and $[h_i, h_j] = 0$ so we have exactly
 $U_1(dt) = \prod_i e^{-ih_i dt} = \prod_i u_i(dt)$

The 2-body part is more tricky

$U_2(t) = e^{-iH_{\nu\nu}t} = e^{-i\sum_{i<j} h_{ij}t}$ where $h_{ij} = J_{ij}\vec{\sigma}_i \cdot \vec{\sigma}_j$ and $[h_{ij}, h_{ik}] \neq 0$.
 We approximate it in pairs
 $U_2(t) \approx \prod_{i<j} e^{-ih_{ij}t} = \prod_{i<j} u_{ij}(t)$
 with an error $\sim \mathcal{O}(t^2)$

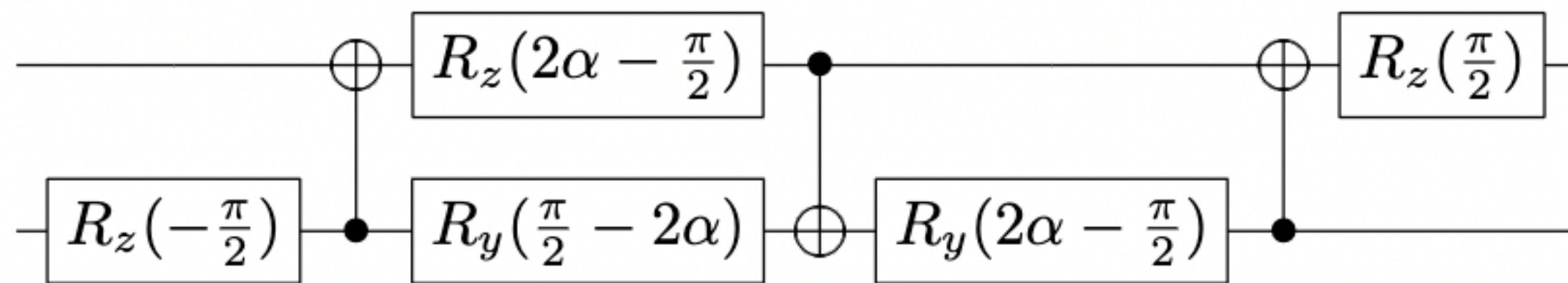
Optimal Circuit for full connectivity



Each pair propagator is simple:

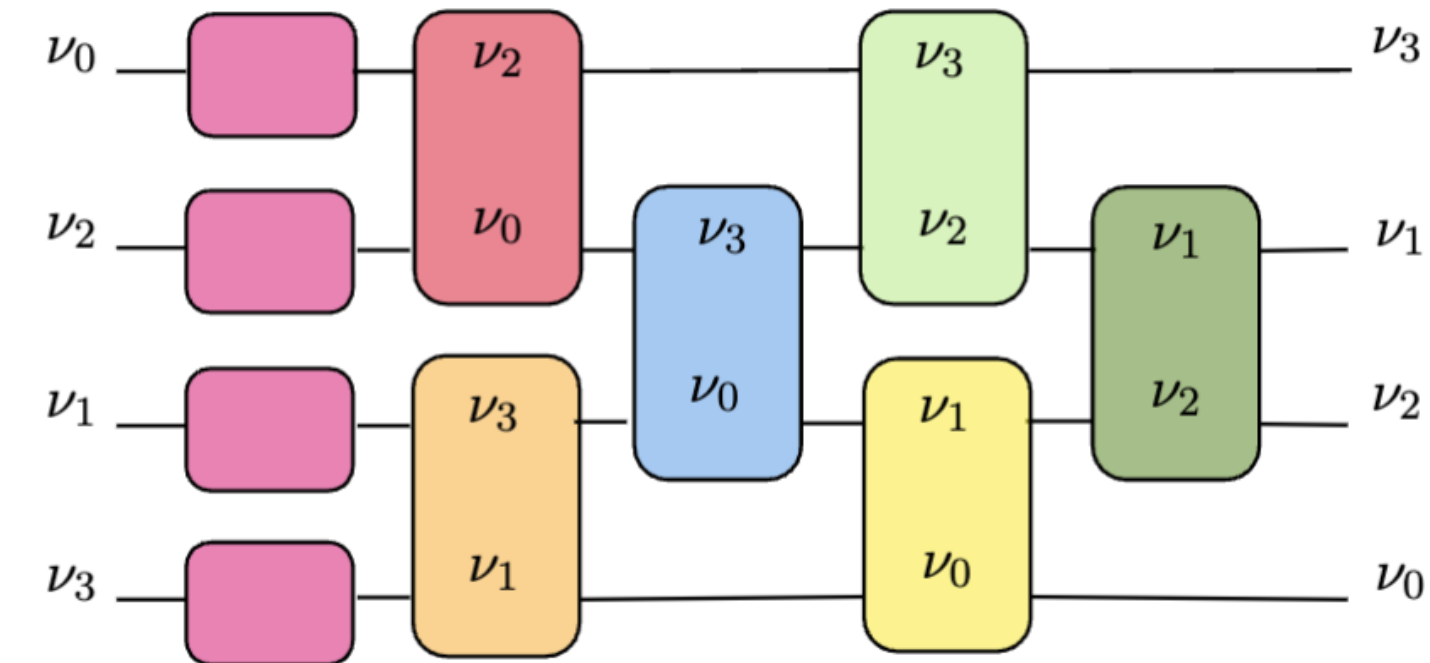
$$u_{ij}(t) = e^{-iJ_{ij}(X_i \otimes X_j + Y_i \otimes Y_j + Z_i \otimes Z_j)t}$$

and has the following optimal CNOT-based circuit where $\alpha = -dtJ_{ij}$



F. Vatan and C. Williams (2004)

Swap network for linear connectivity

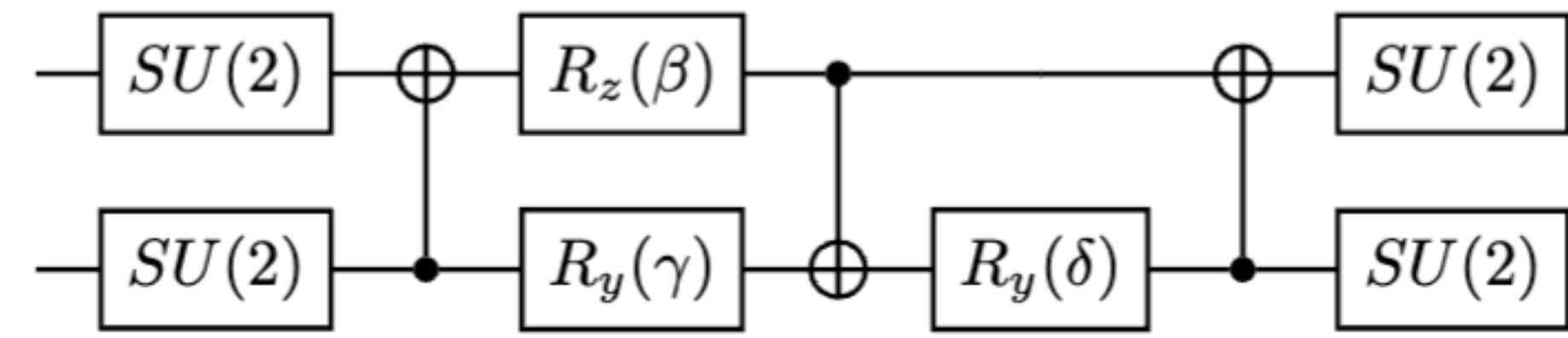


Hall, A. Roggero et. al (2021)

Each pair propagator contains also a SWAP operation:

$$w_{ij}(t) = \text{SWAP}_{ij} u_{ij}(t)$$

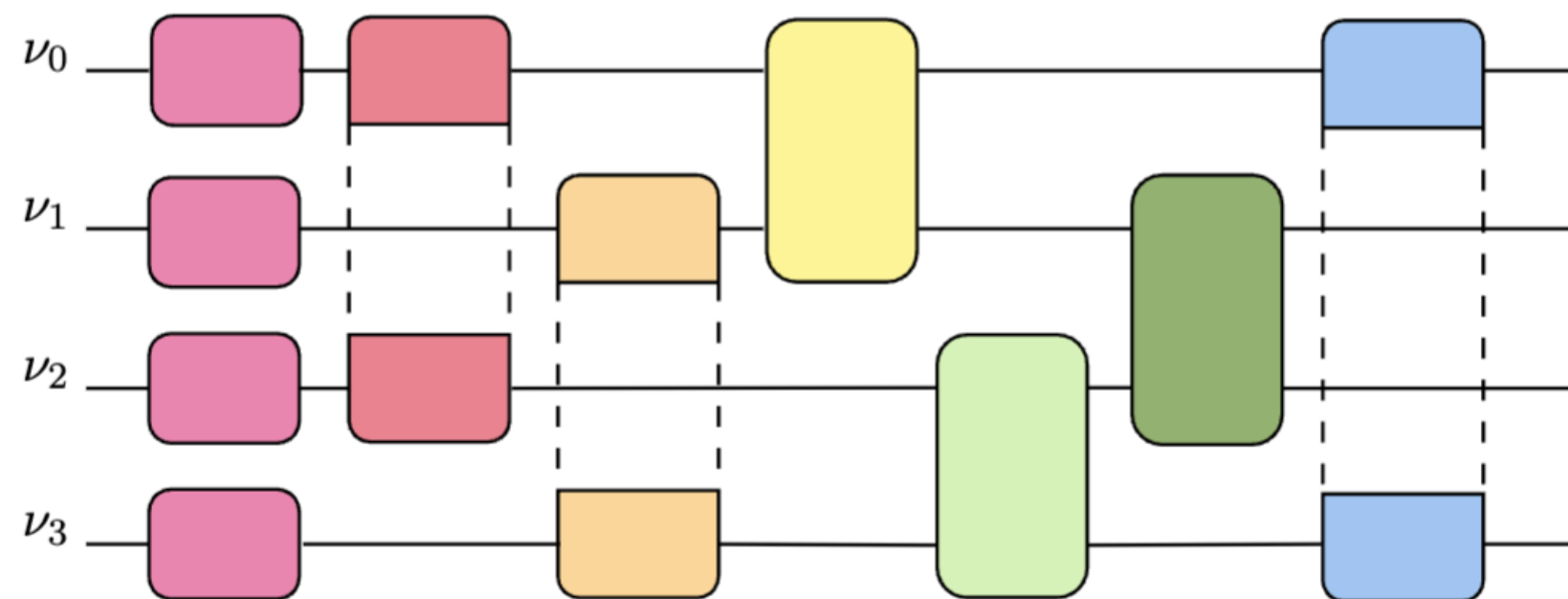
And it requires more single qubit rotations



F. Vatan and C. Williams (2004)

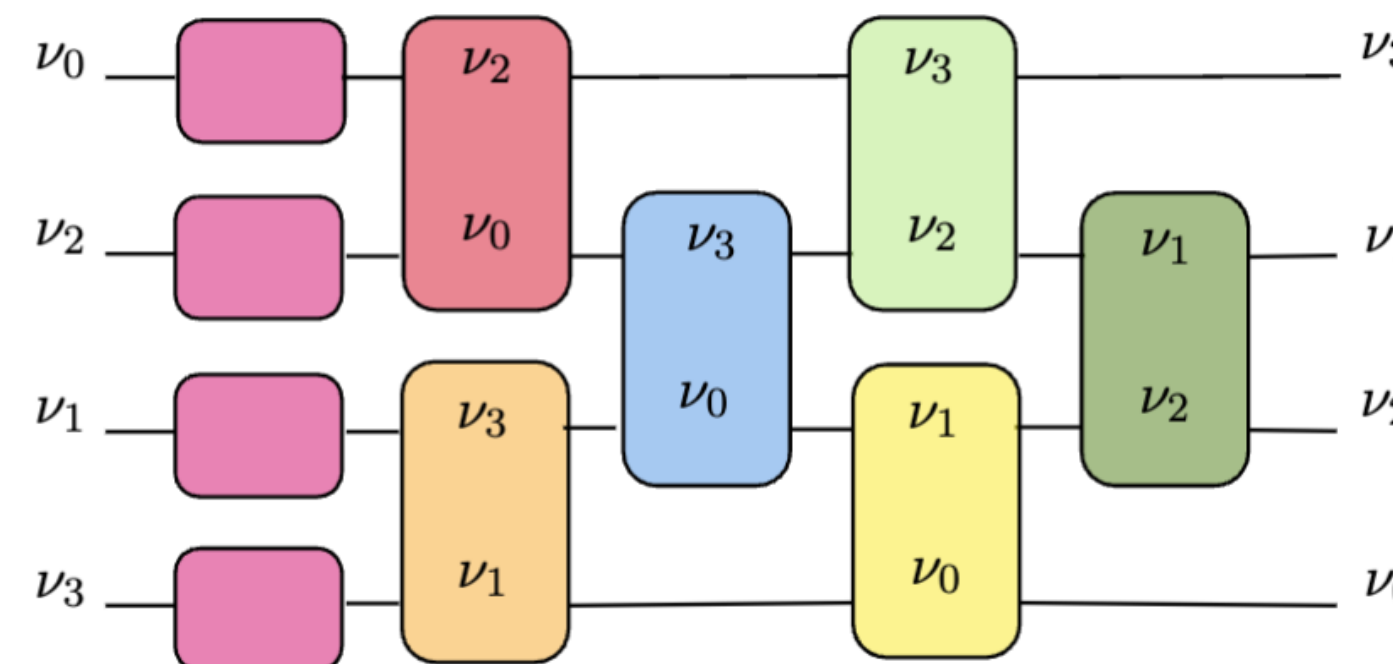
QUBIT CONNECTIVITY: TROTTER ERROR

Optimal Circuit for full connectivity



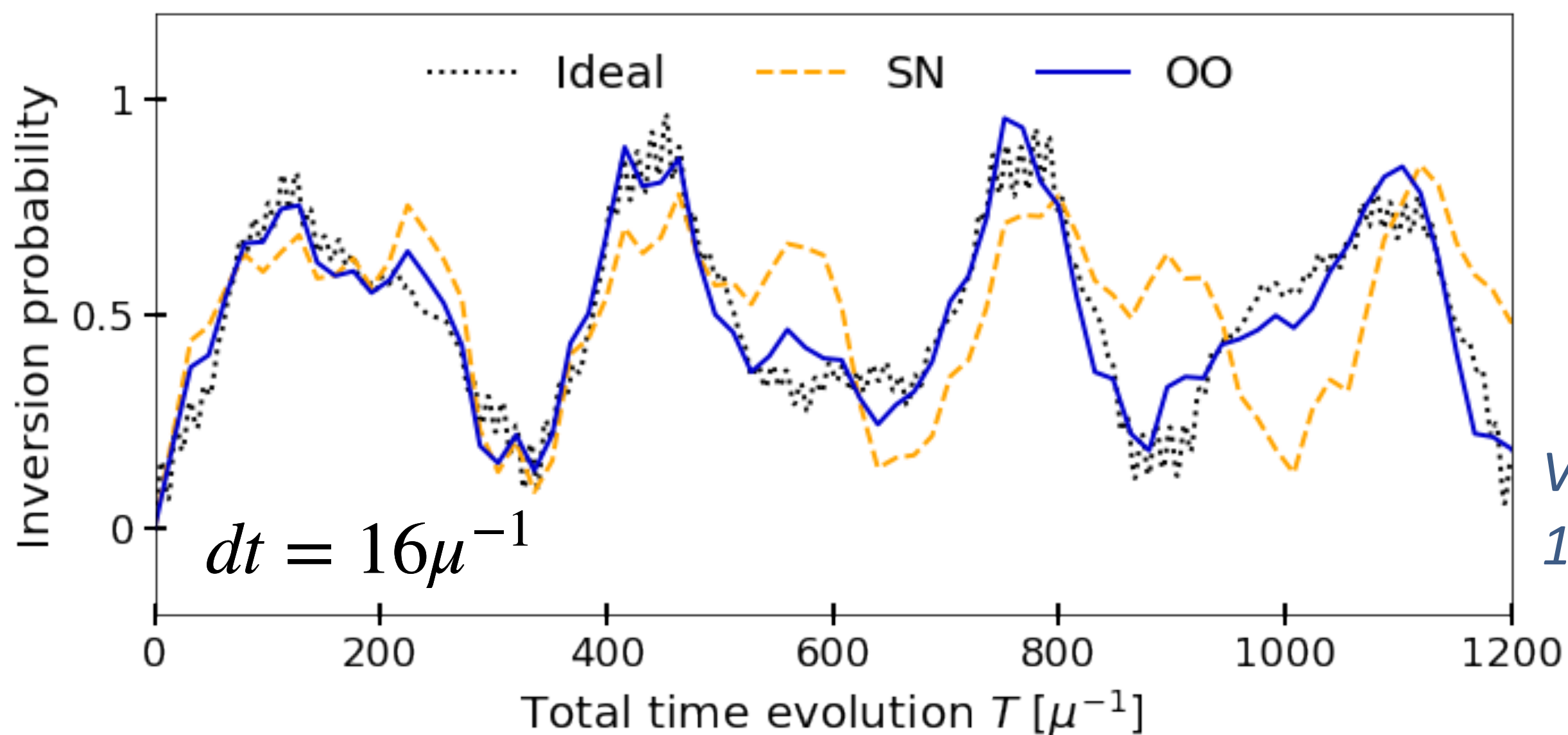
Full freedom in the pair ordering

Swap network for linear connectivity



Hall, A. Roggero et. al (2021)

Full freedom in the initial encoding, but the pair ordering is fixed. Any order can be obtained at the price of adding additional SWAP gates.



V. Amitrano et. al. Phys. Rev. D 107, 023007 (2023)

THE UNITARY IMPLEMENTATION: MACHINE AWARE COMPILATION

Quantinum System Model (QSM) H1-2

- Trapped-ion device
- Full-connected qubits
- High fidelity: $\varepsilon_q \sim 10^{-4}$ and $\varepsilon_{qq} \sim 10^{-3}$

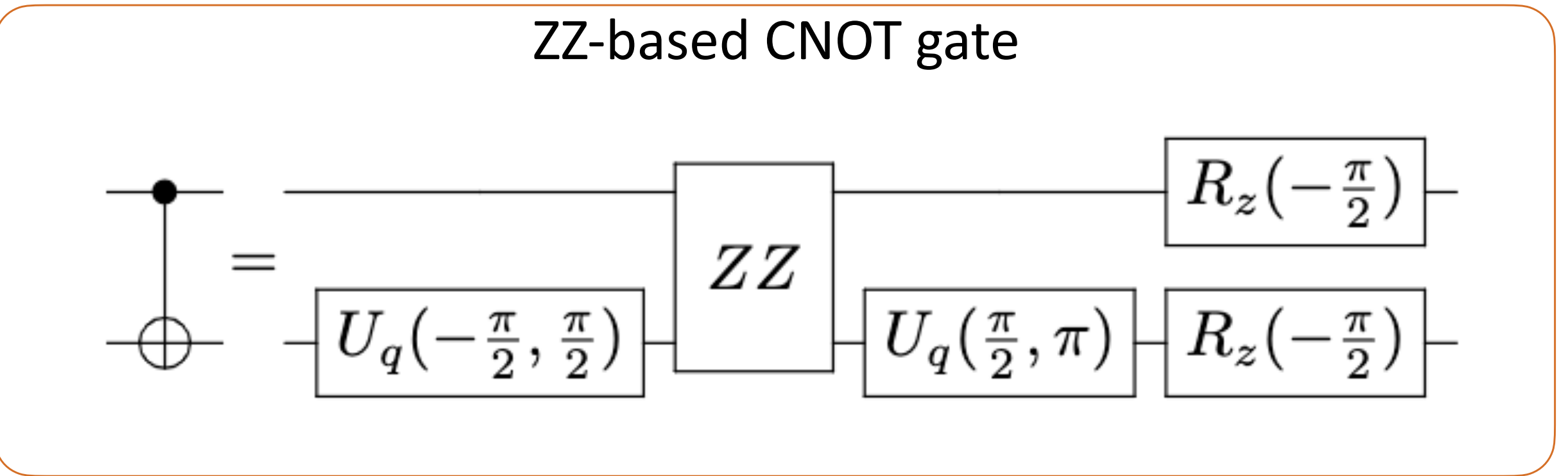
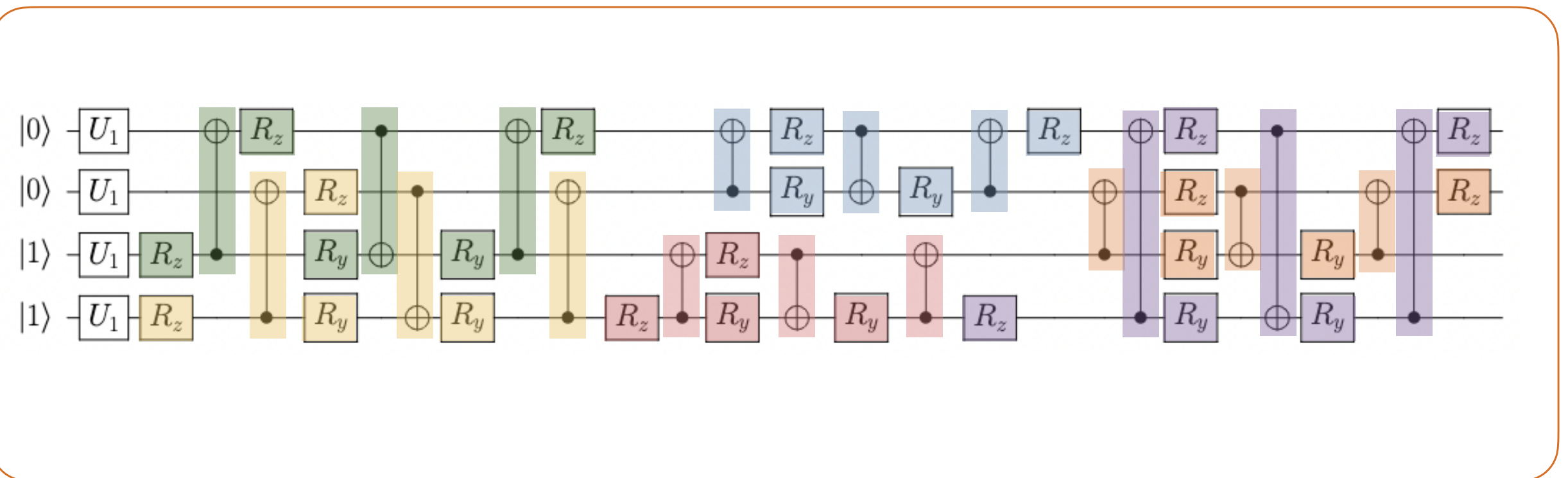
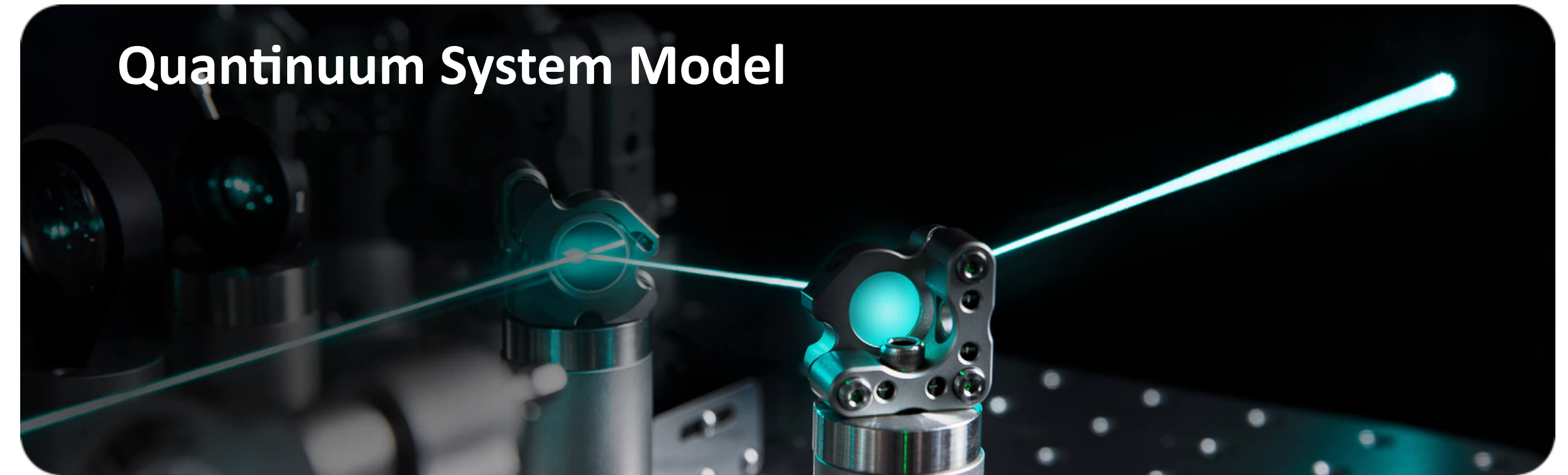
Machine aware compilation:

- Qubit topology
- Quantum gate set

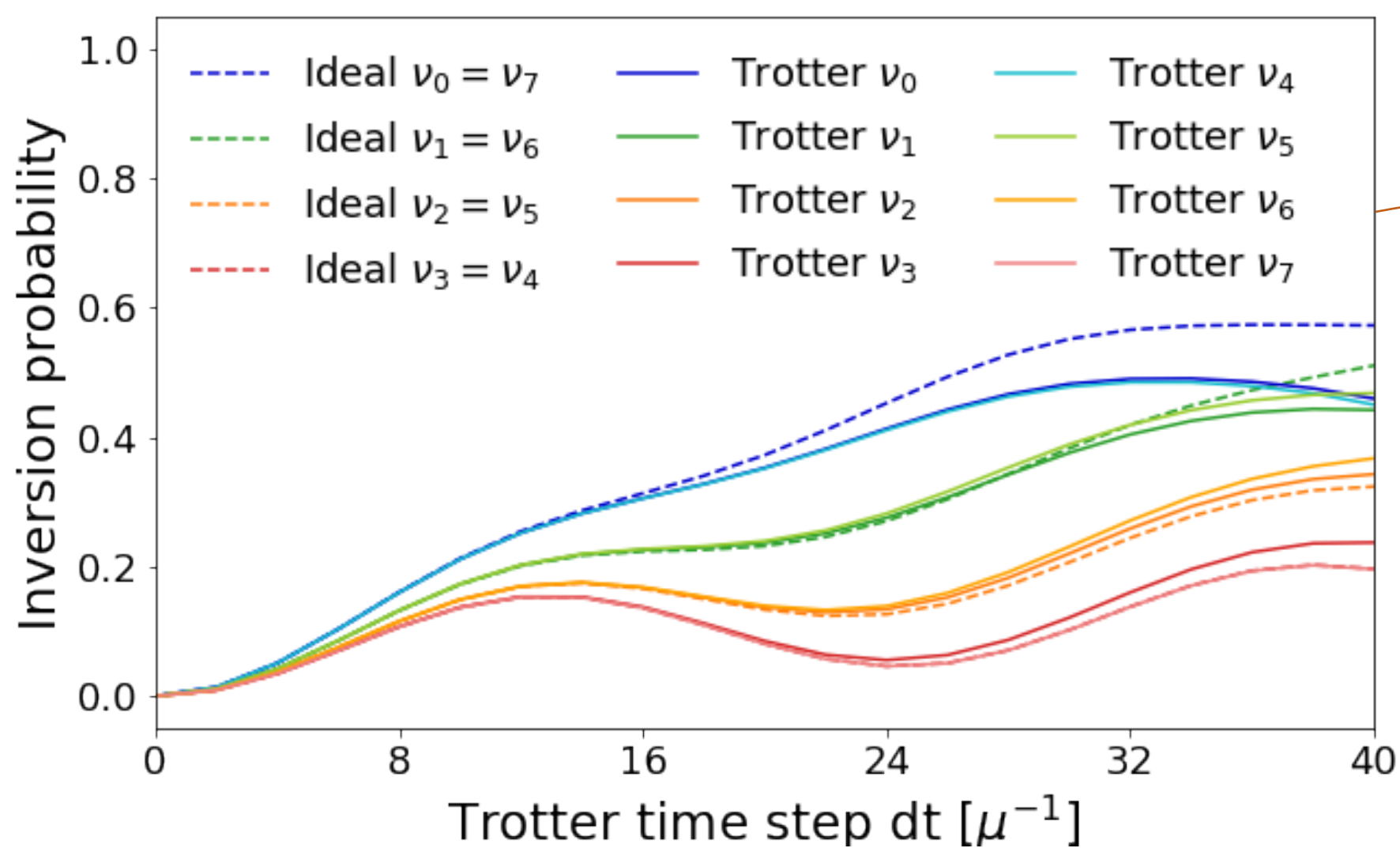
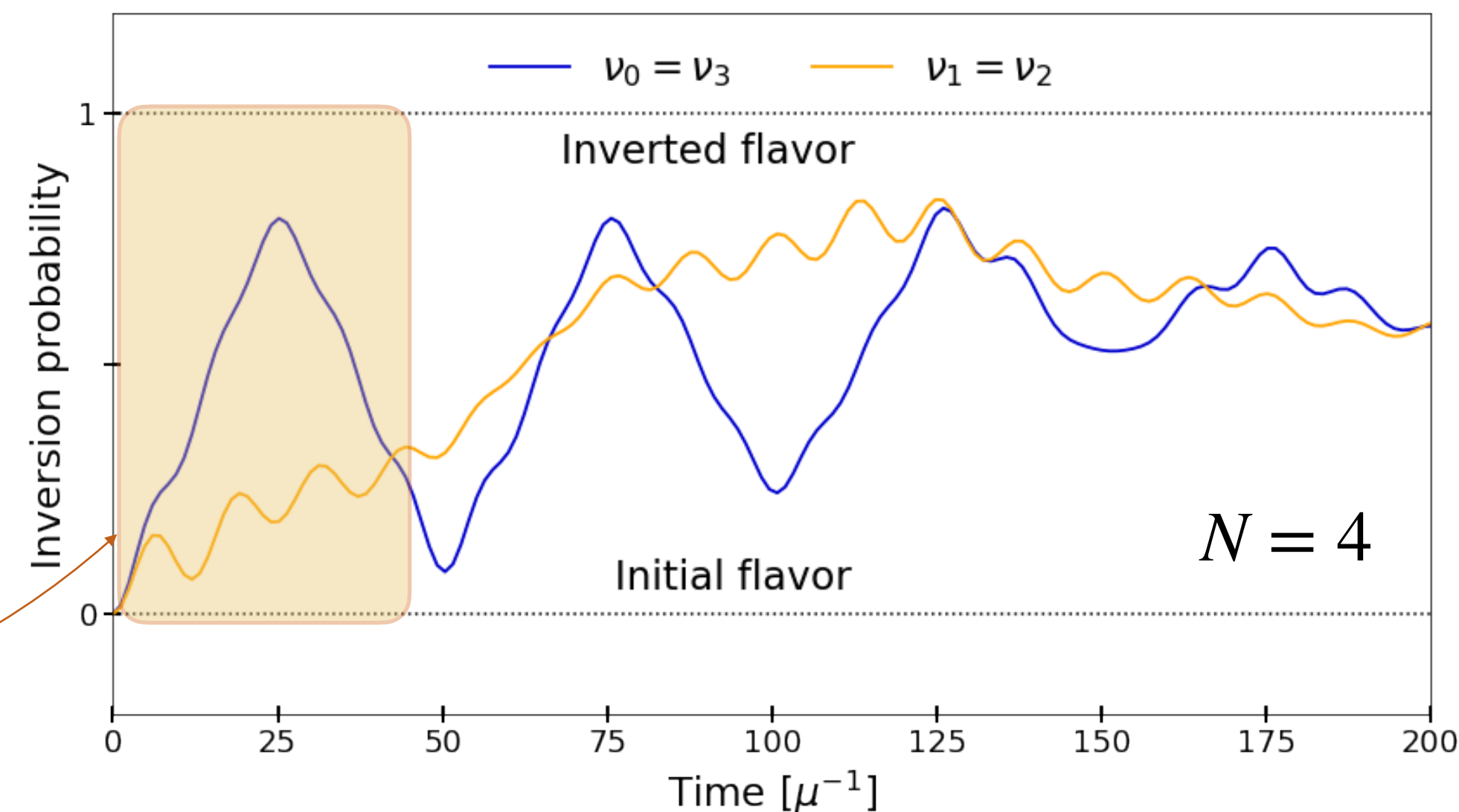
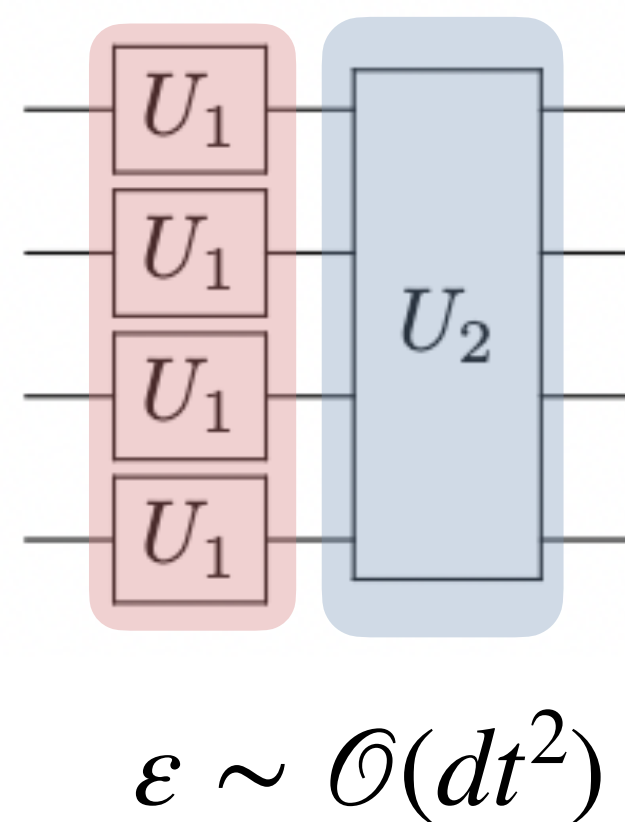
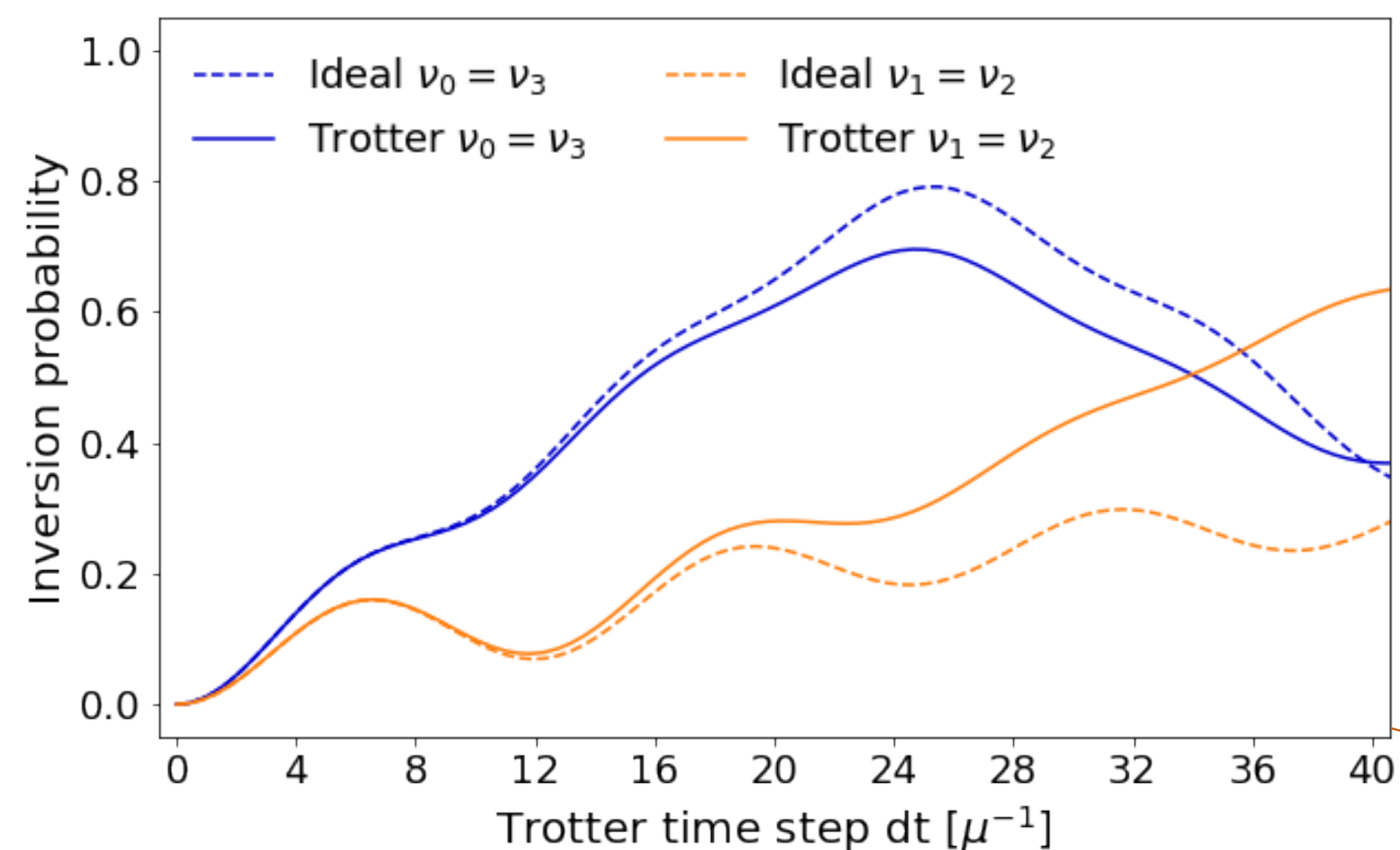
- $$R_z(\lambda) = \begin{pmatrix} e^{-i\lambda/2} & 0 \\ 0 & e^{i\lambda/2} \end{pmatrix}$$

- $$U_q(\theta, \varphi) = \begin{pmatrix} \cos \theta/2 & -ie^{-i\varphi} \sin \theta/2 \\ -ie^{i\varphi} \sin \theta/2 & \cos \theta/2 \end{pmatrix}$$

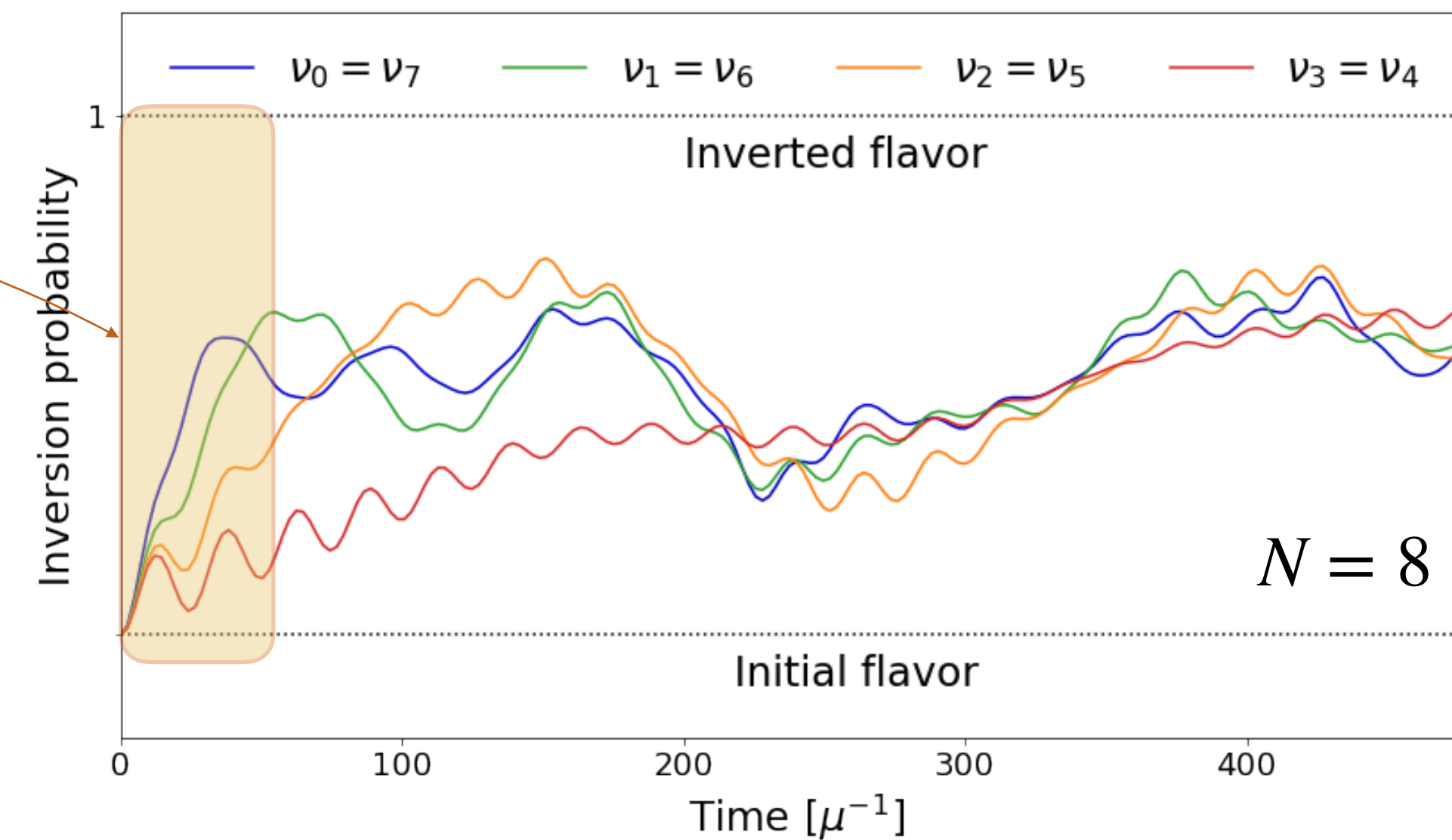
- $$ZZ = e^{-i\frac{\pi}{4}Z \otimes Z} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & i & 0 & 0 \\ 0 & 0 & i & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$



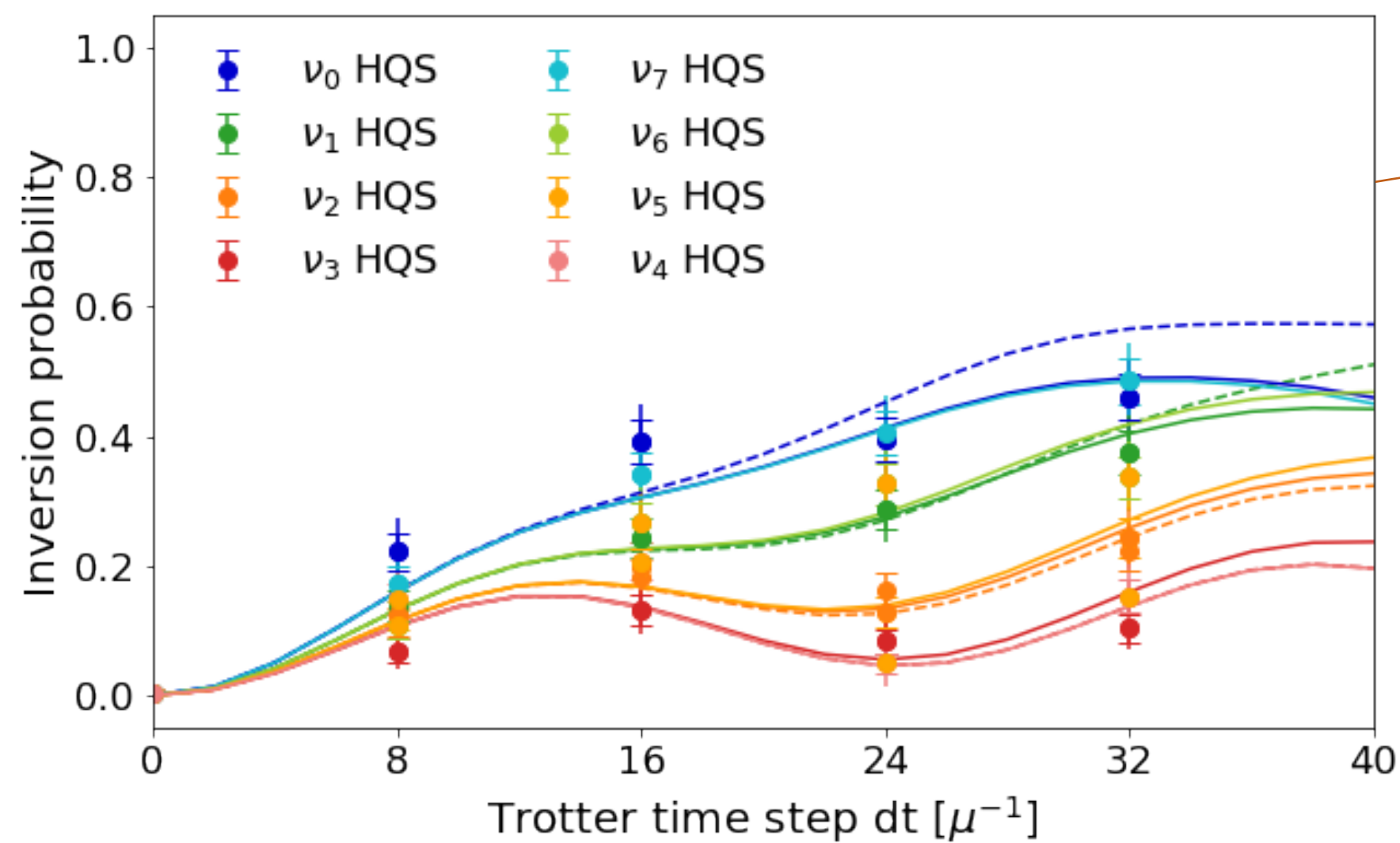
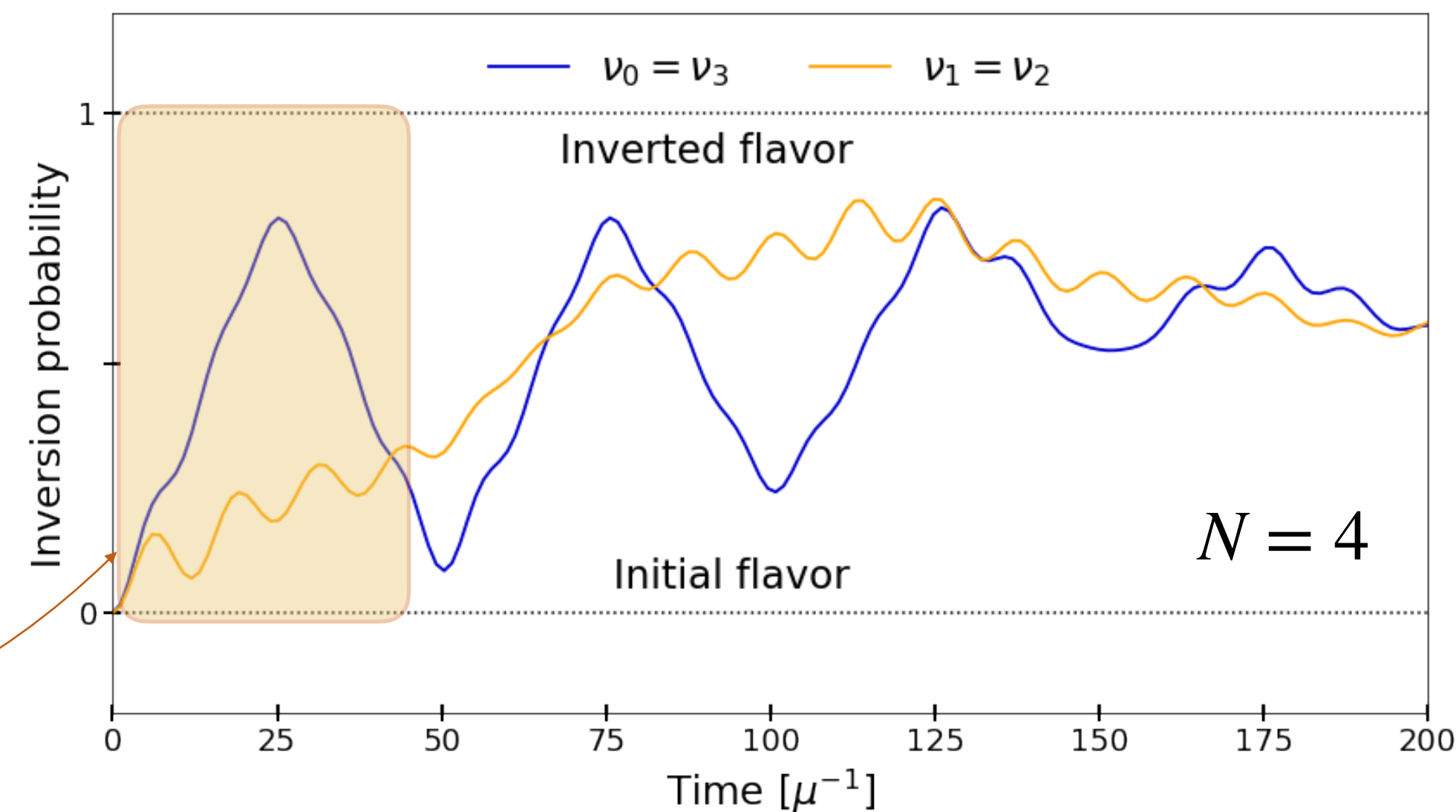
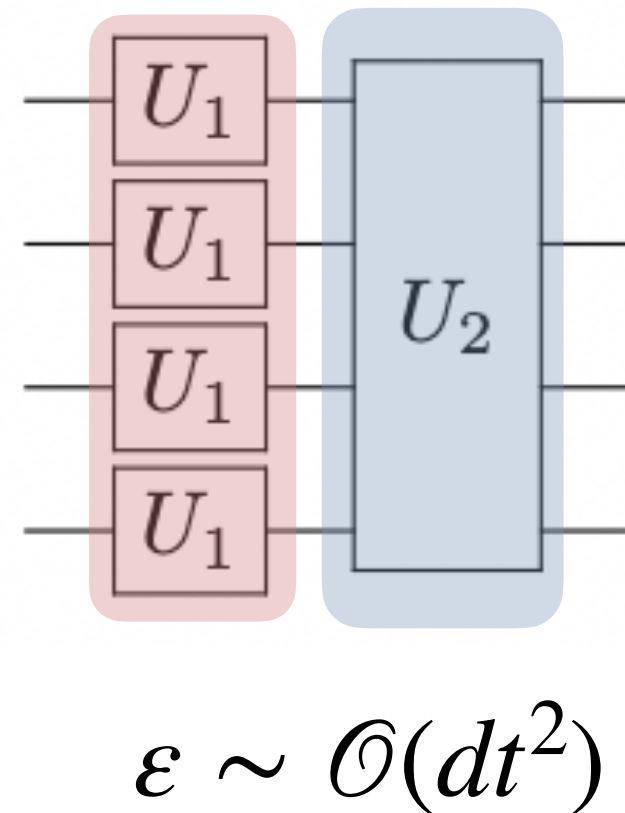
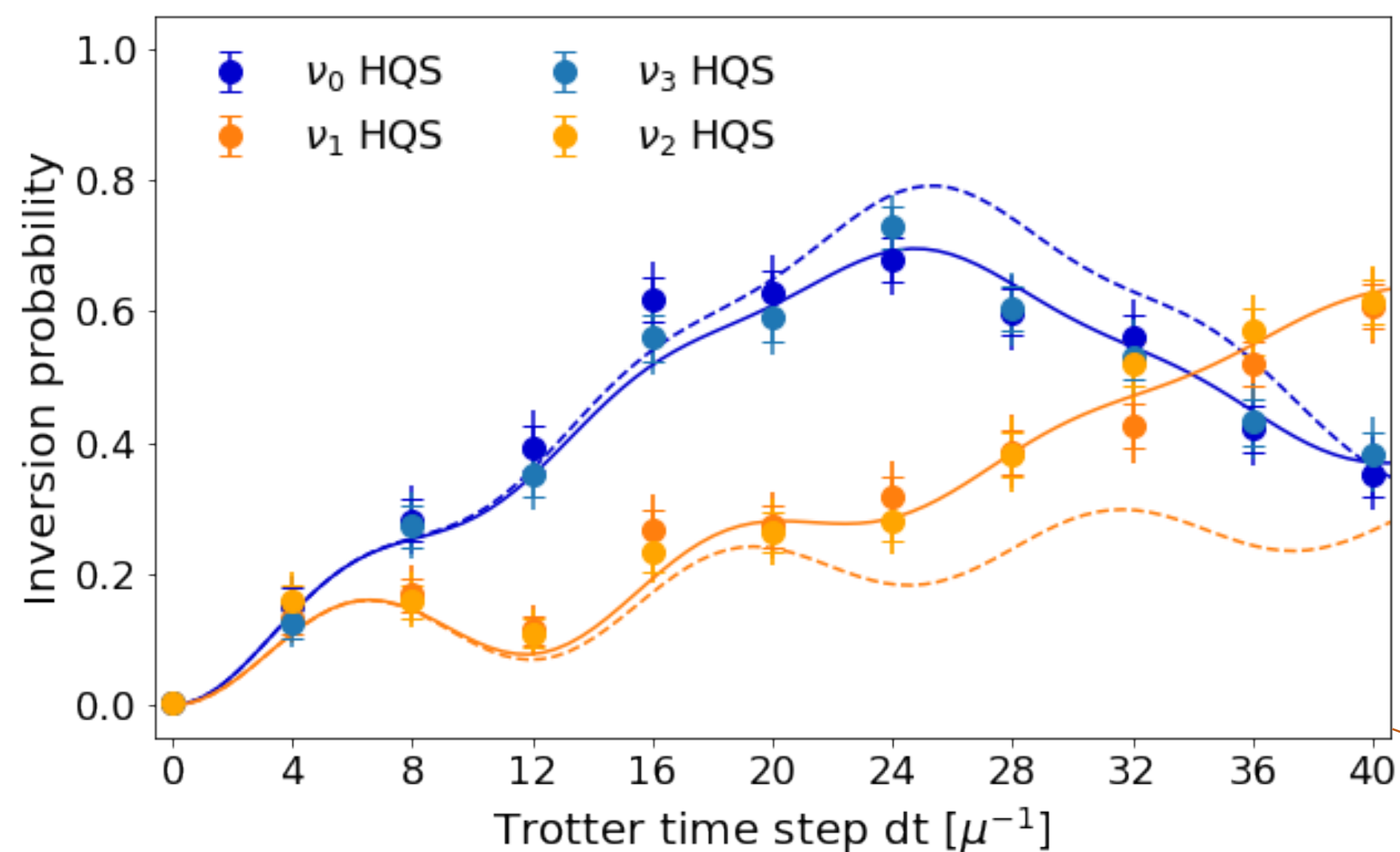
RESULTS: SINGLE TROTTER STEP



V. Amitrano et. al.
Phys. Rev. D 107,
023007 (2023)

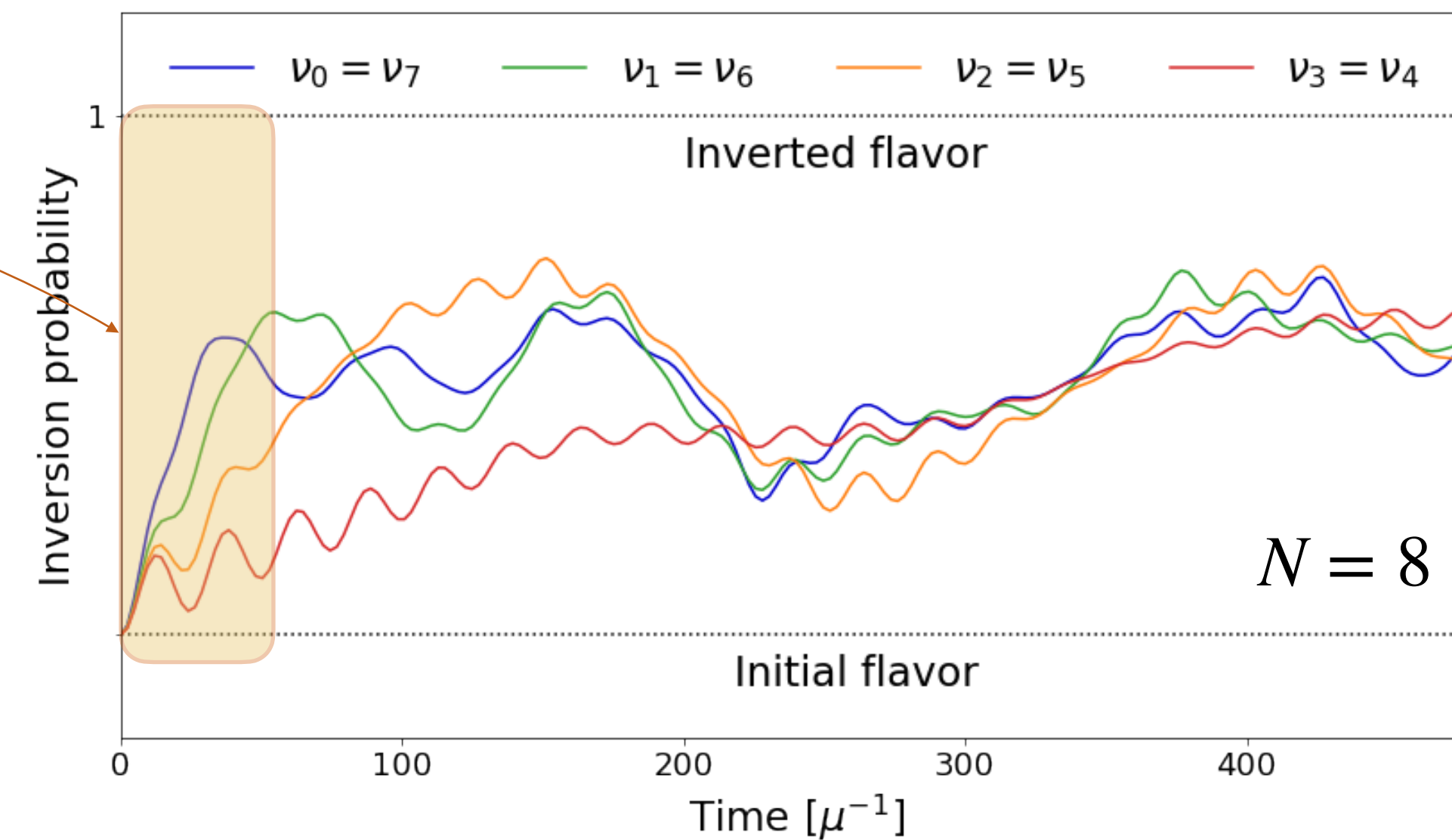


RESULTS: SINGLE TROTTER STEP

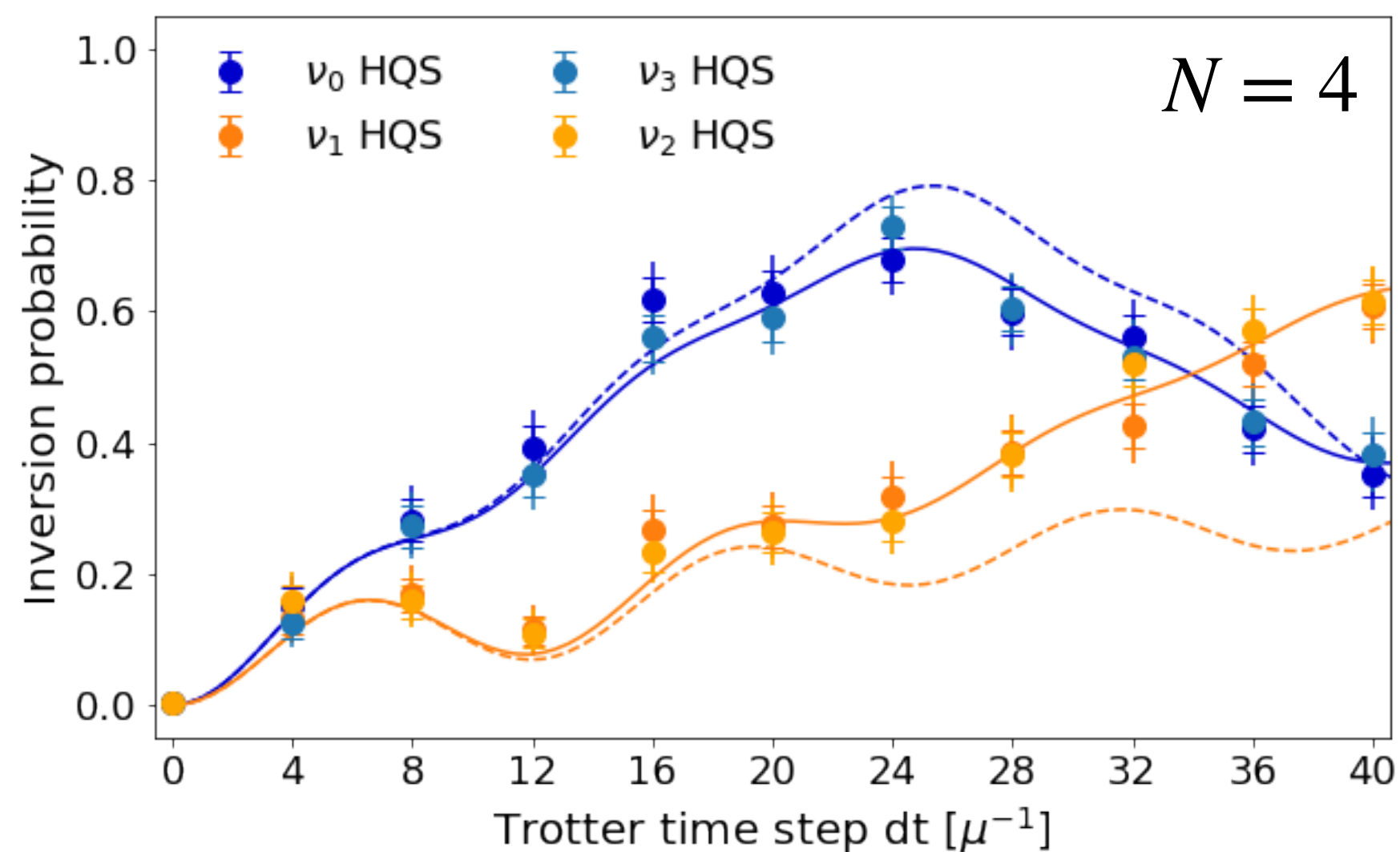


error bar = 68 %

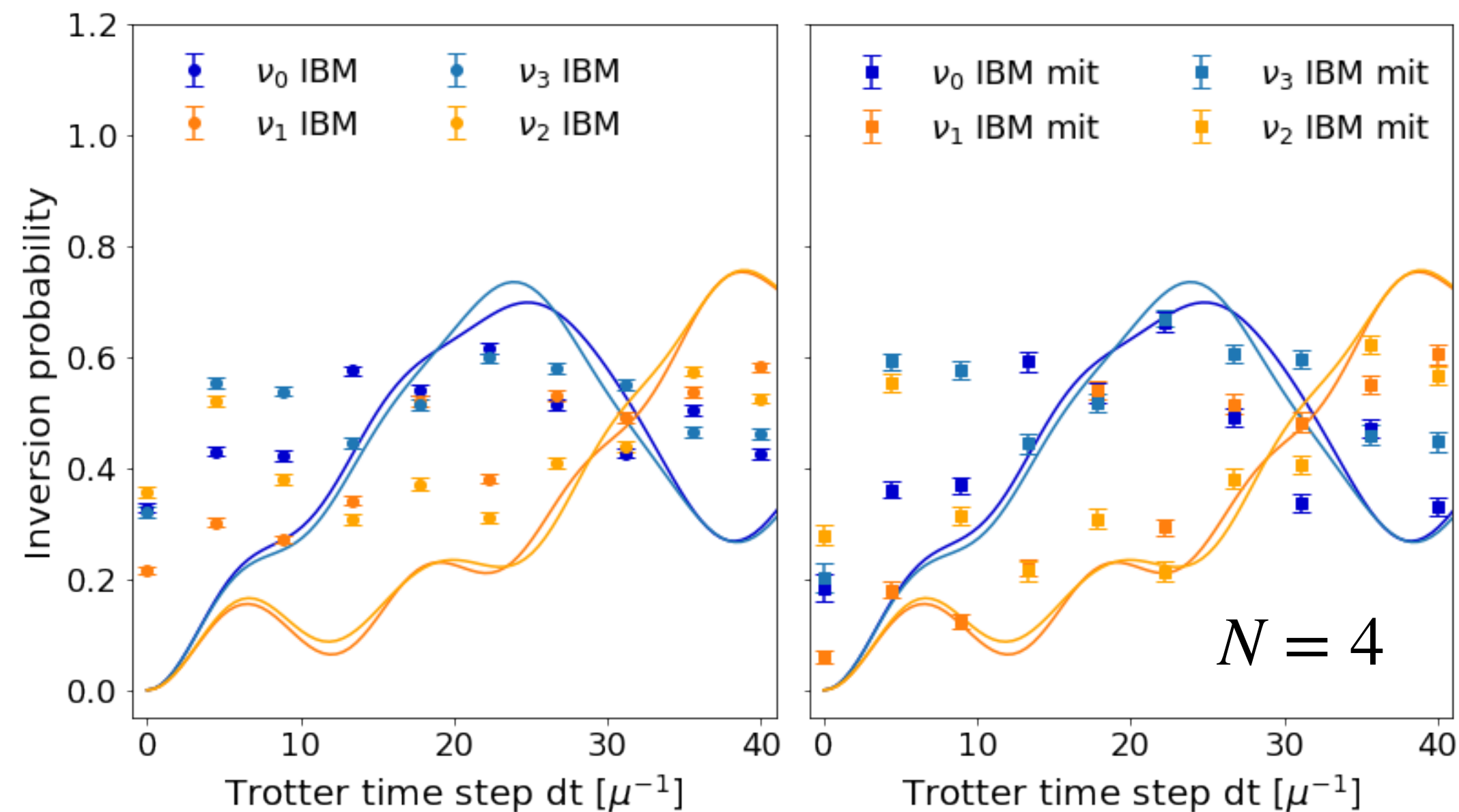
V. Amitrano et. al.
Phys. Rev. D 107,
023007 (2023)



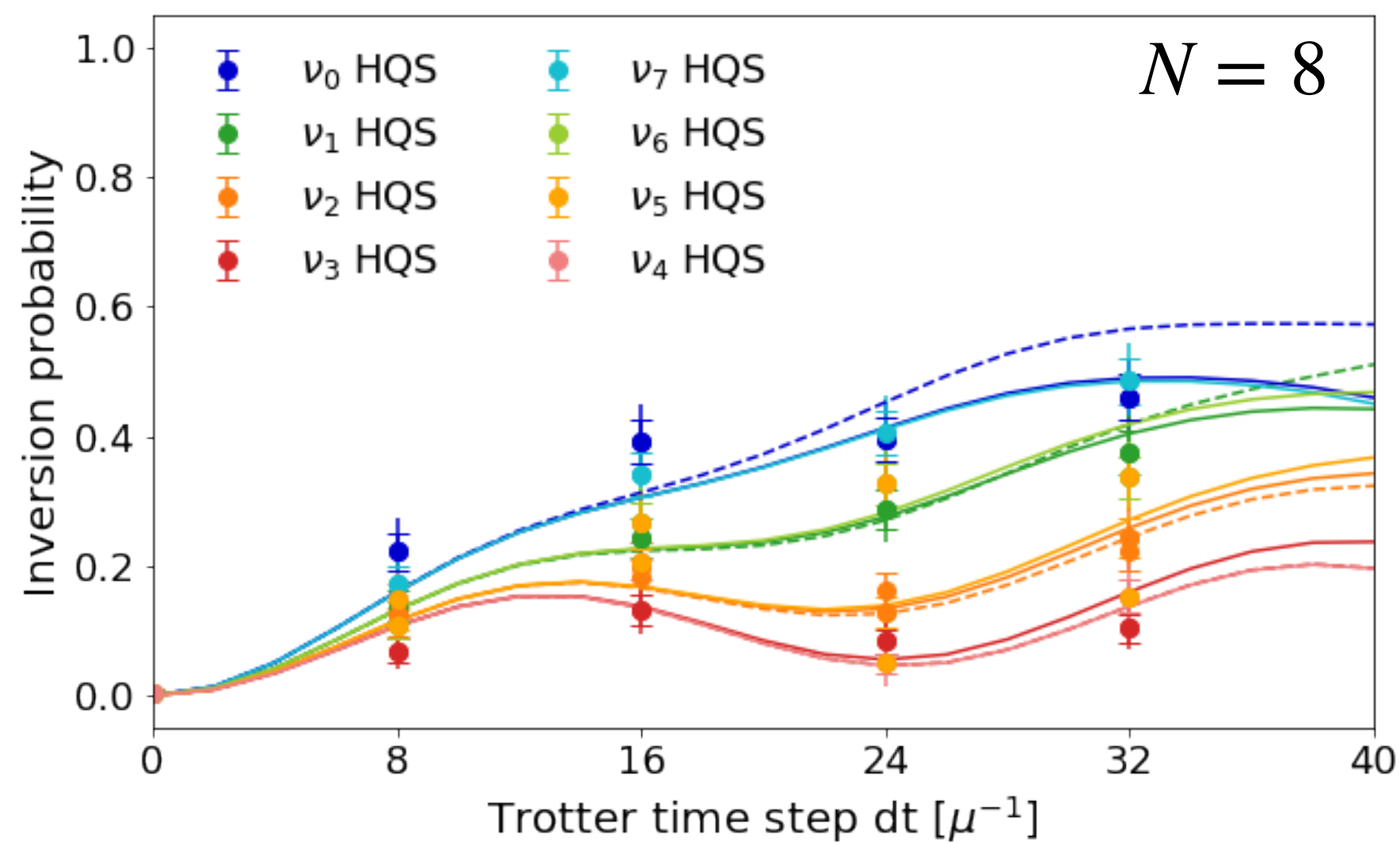
RESULTS: SINGLE TROTTER STEP



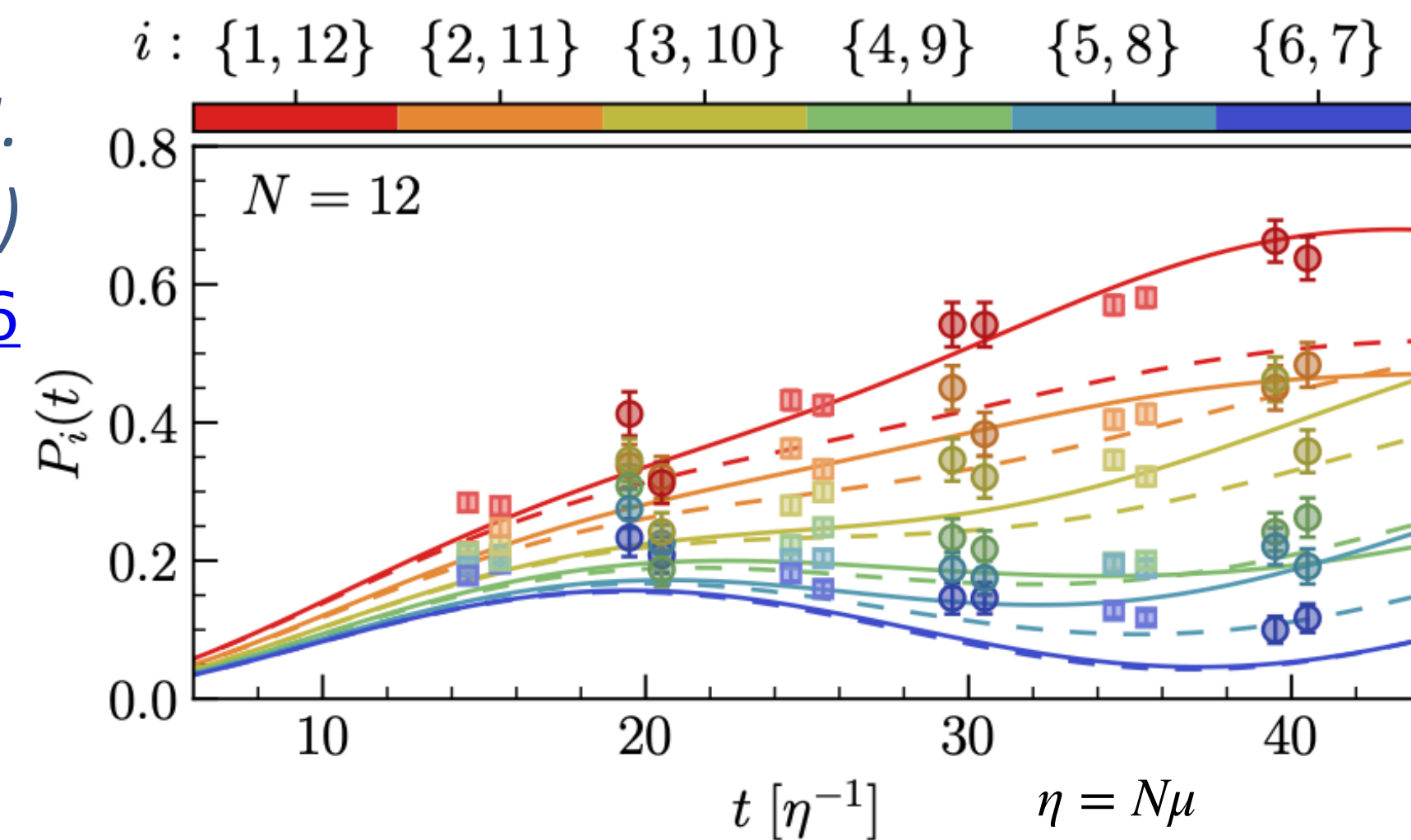
B. Hall et. al. Phys. Rev. D 104, 063009 (2021)



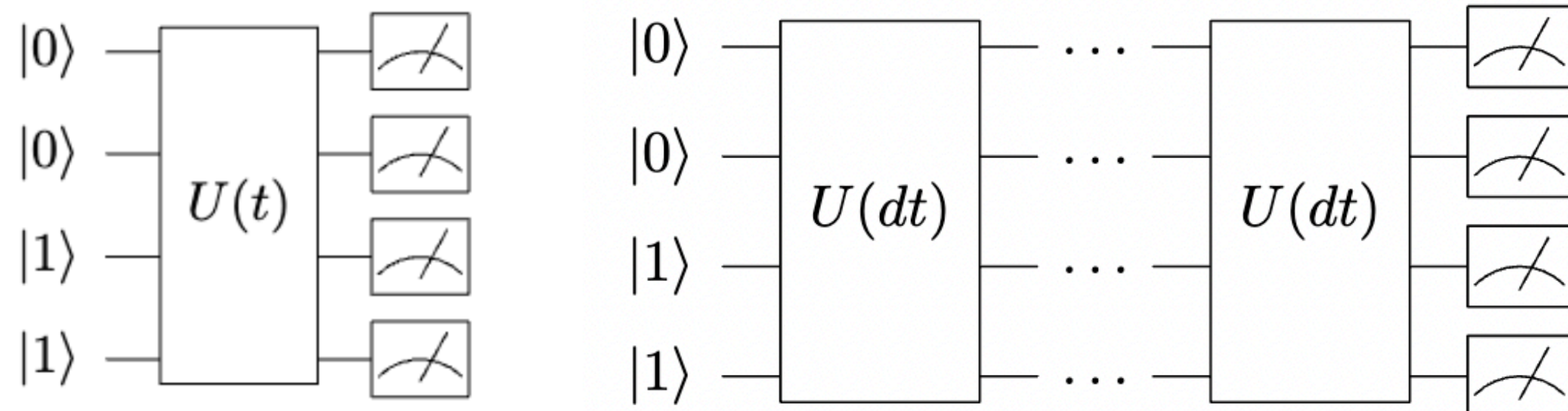
M. Illa and M. J. Savage (2022)
[arXiv:2210.08656](https://arxiv.org/abs/2210.08656)



V. Amitrano et. al. Phys. Rev. D 107, 023007 (2023)



RESULTS: MULTIPLE TROTTER STEPS

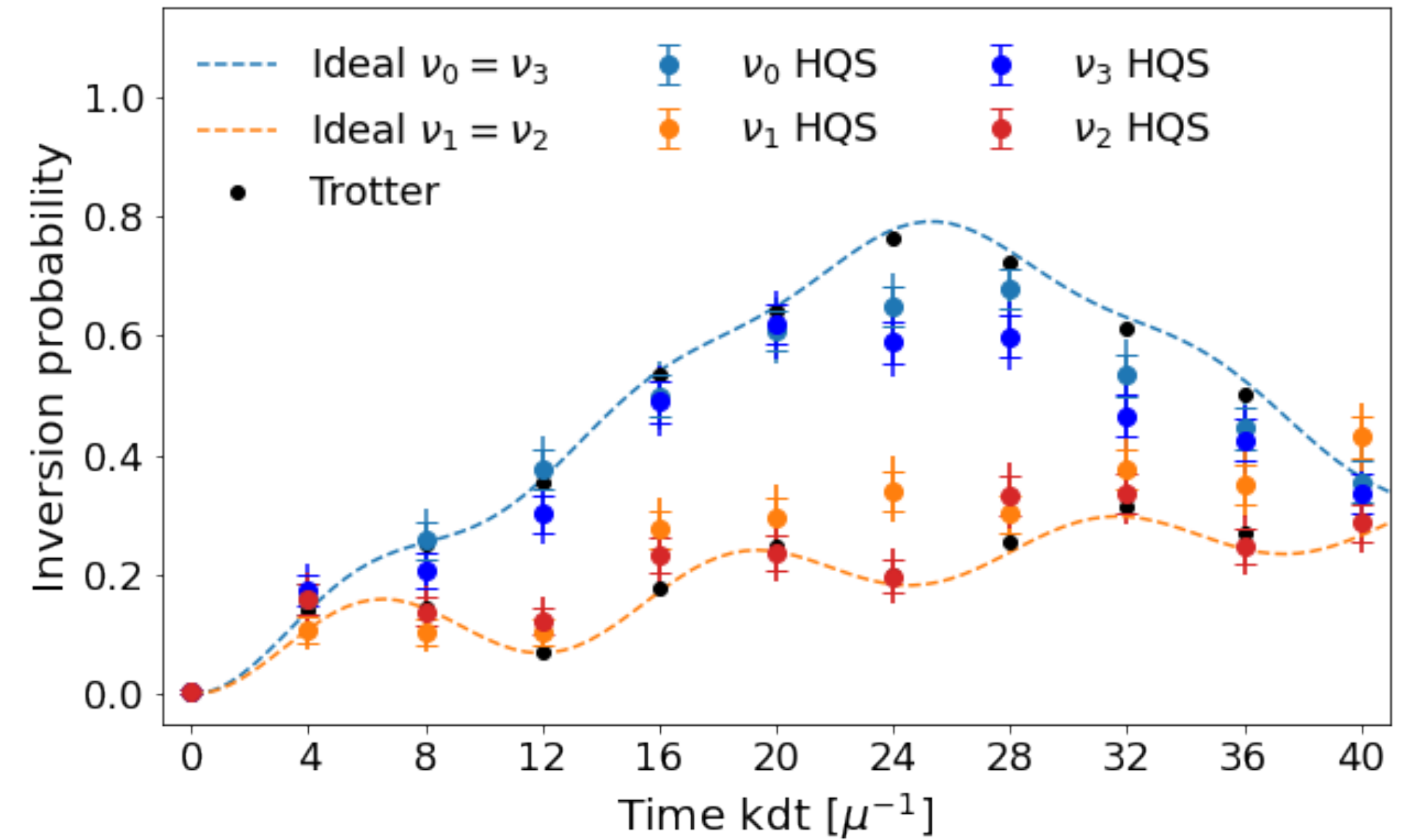
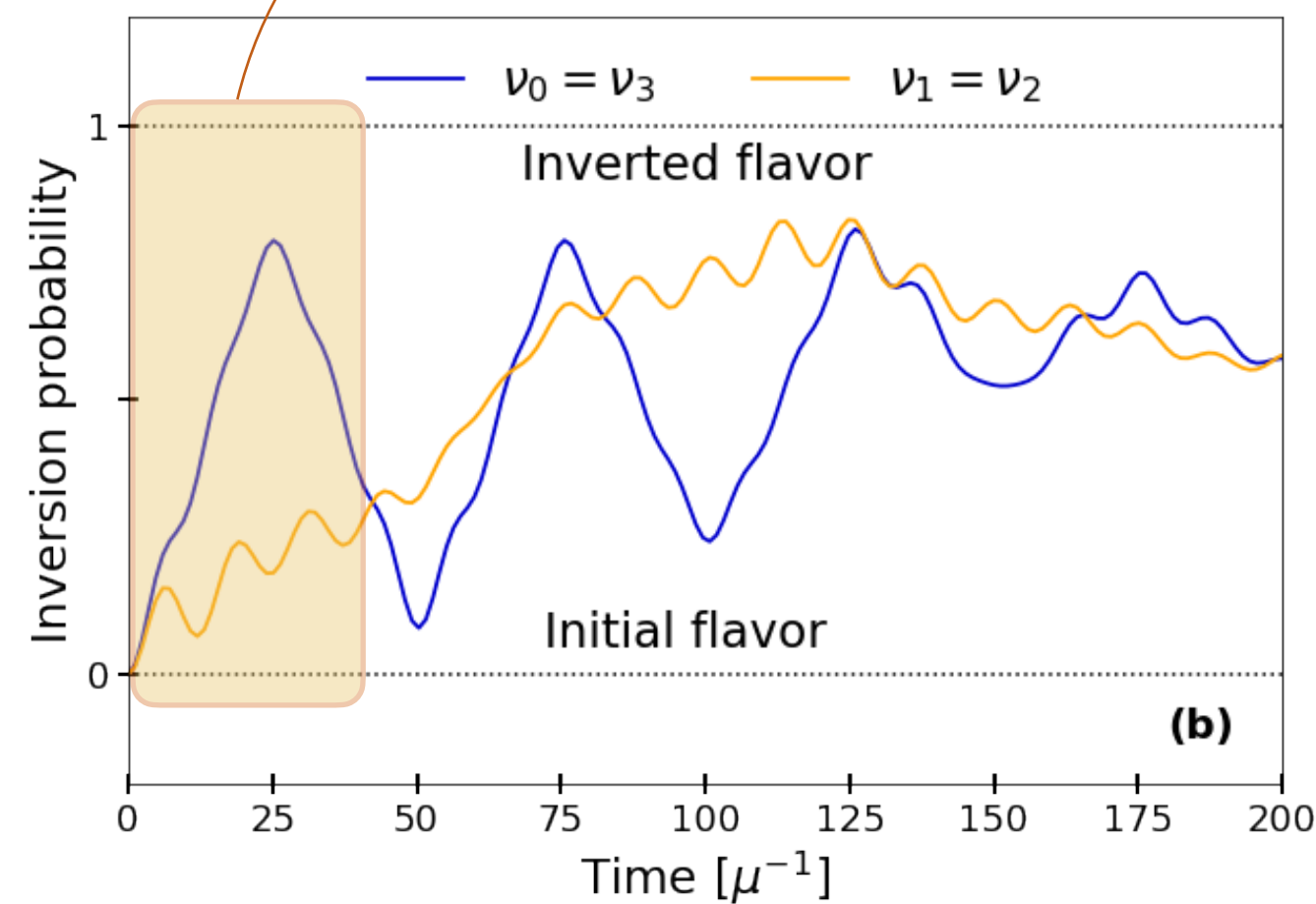


$$\varepsilon \sim \mathcal{O}(k^2 dt^2)$$

$$t = kdt$$

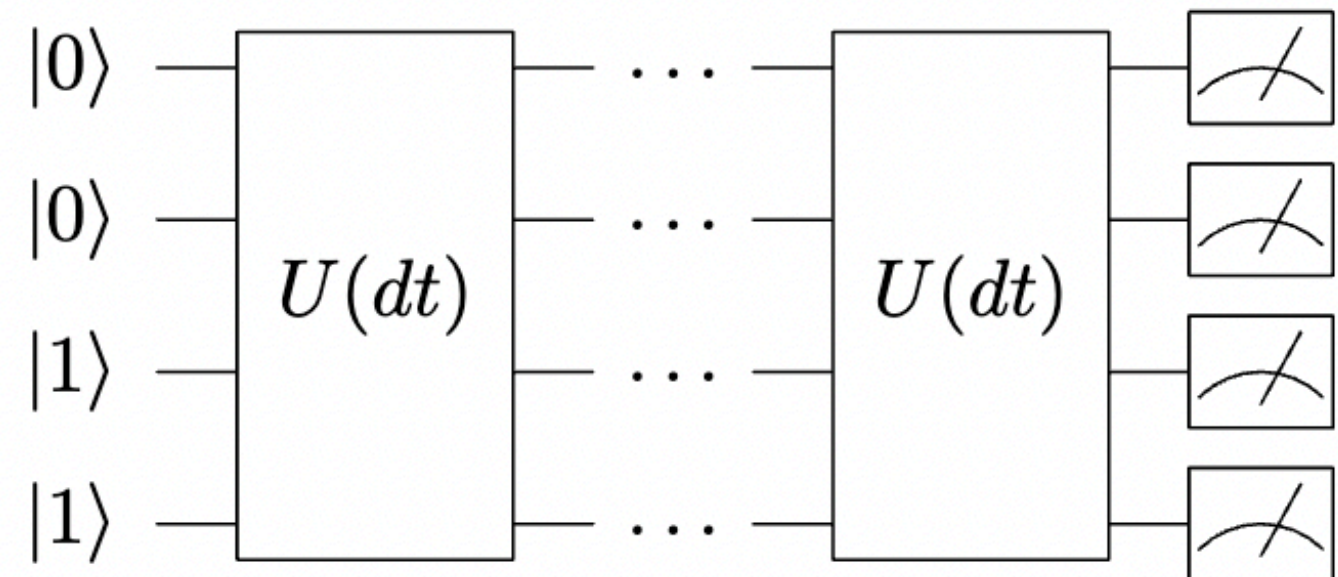
$$\varepsilon \sim \mathcal{O}(kdt^2)$$

- Short time-step $dt = 4\mu^{-1}$
- Ideal \approx trotterized evolution
- Very long quantum circuits (noise)



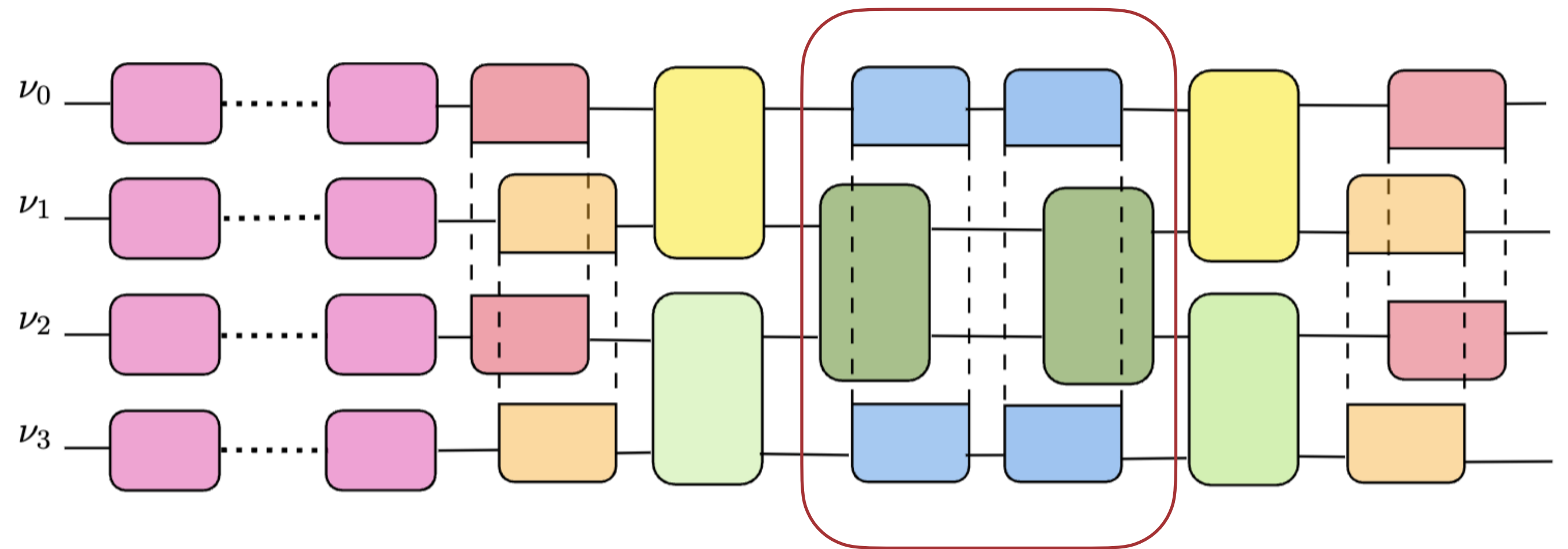
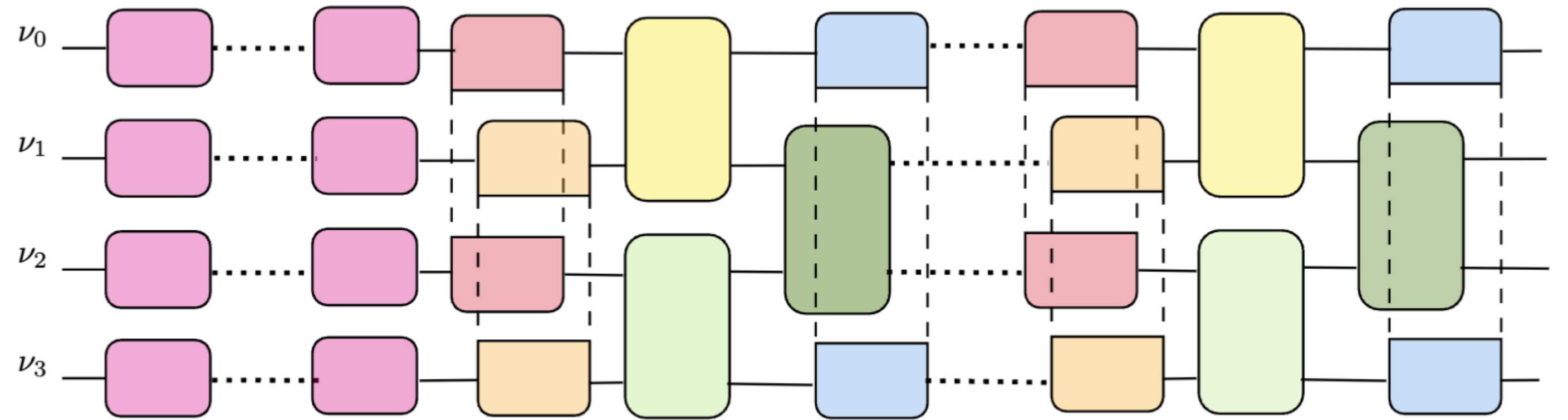
Steps	1	2	3	4	5	6	7	8	9	10
# ZZ	18	36	54	72	90	108	126	144	162	180
# SU(2)	36	68	100	132	164	196	228	260	292	324

INVERSE ORDER FOR THE TROTTER DECOMPOSITION



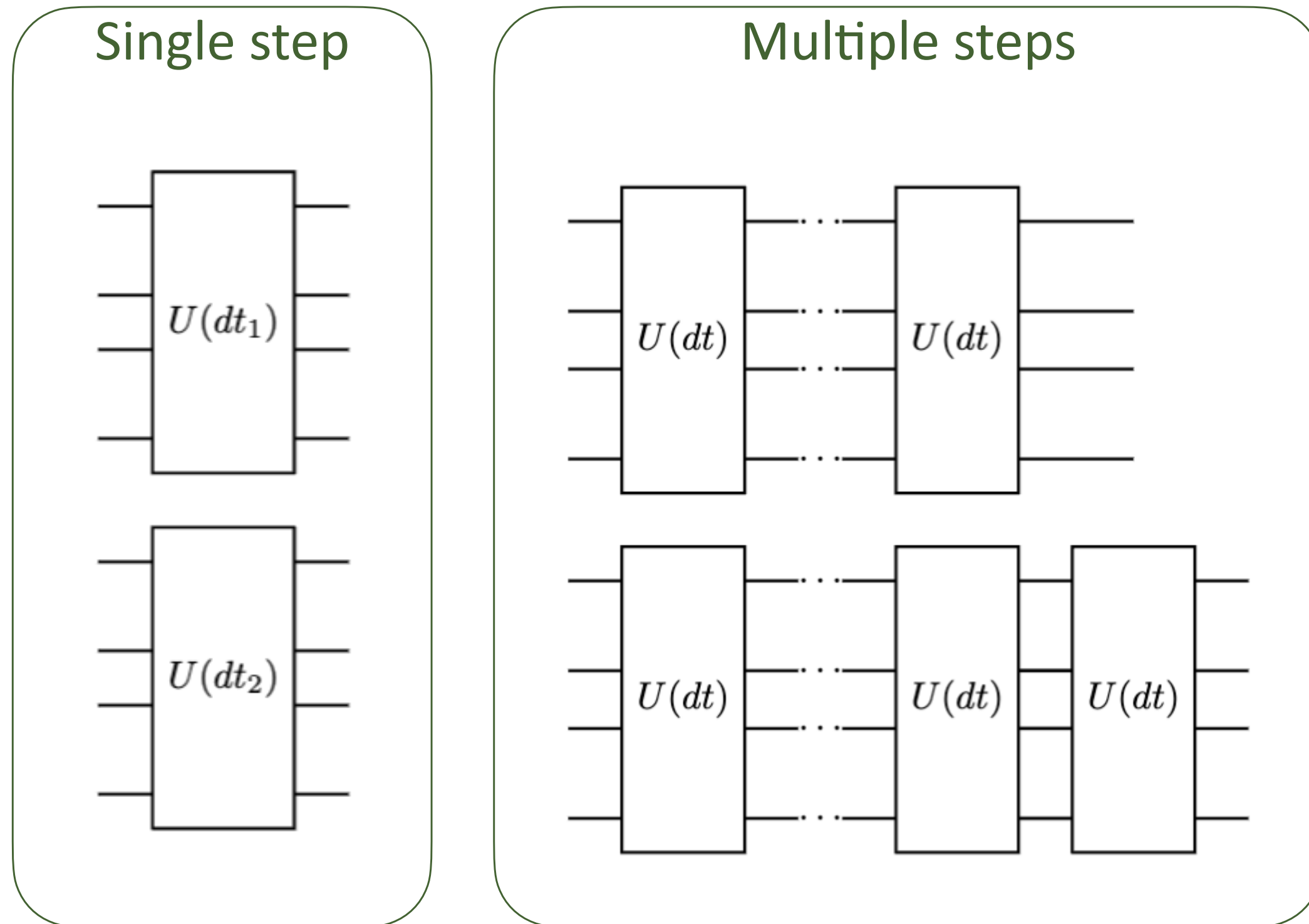
- Alternate steps with inverted order
- Is effectively a second order Trotter decomposition
- For r steps it decreases the number of two-qubit operation of $6r$

$$3\frac{N}{2}r \text{ in general}$$

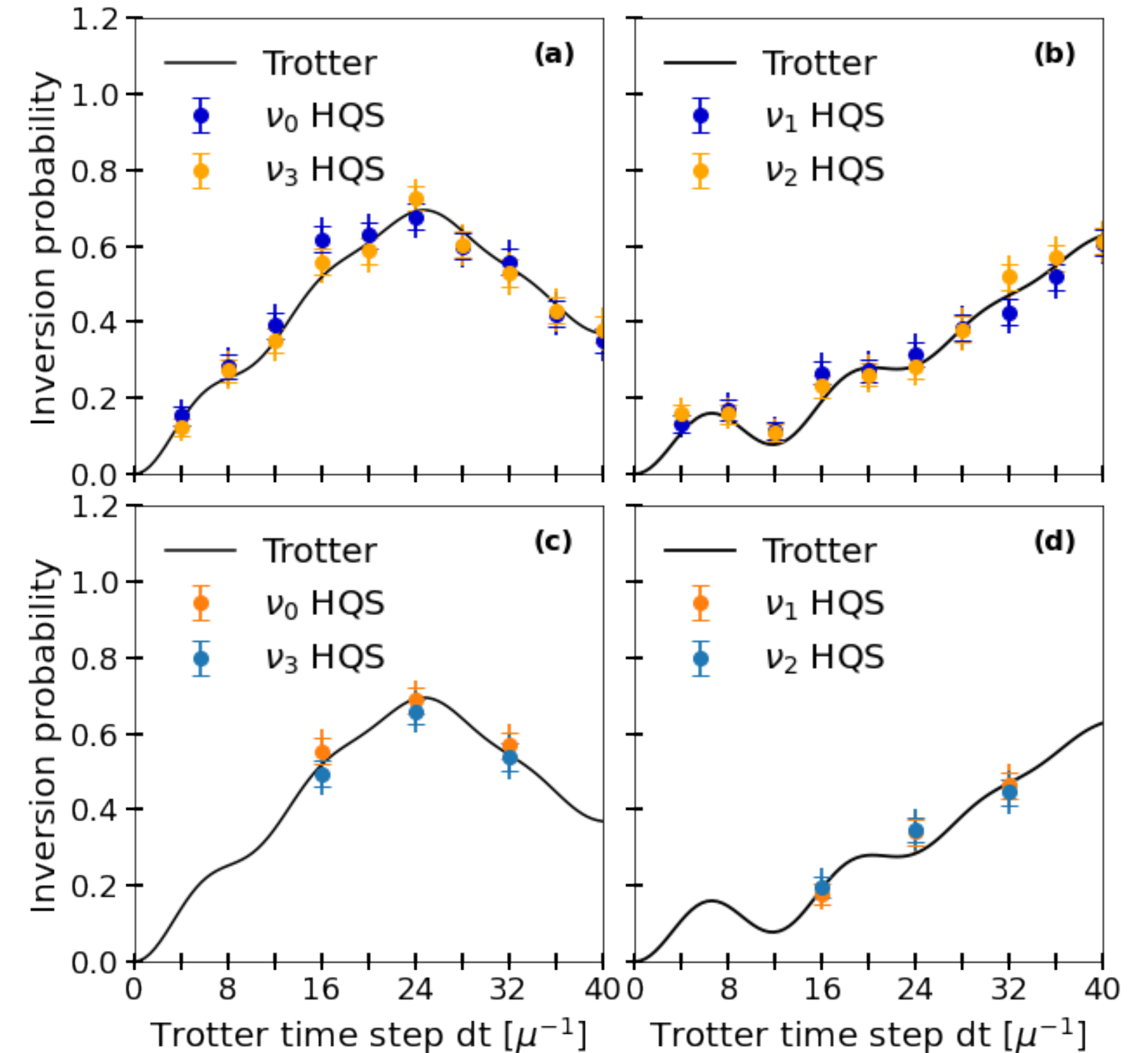


CROSS TALK BETWEEN DISCONNECTED REGISTERS

- All results were obtained running two parallel quantum circuits



- We run 3 parallel circuits to check the cross talk effect
- The results are compatible with the previous one:
 - Cross talk is negligible

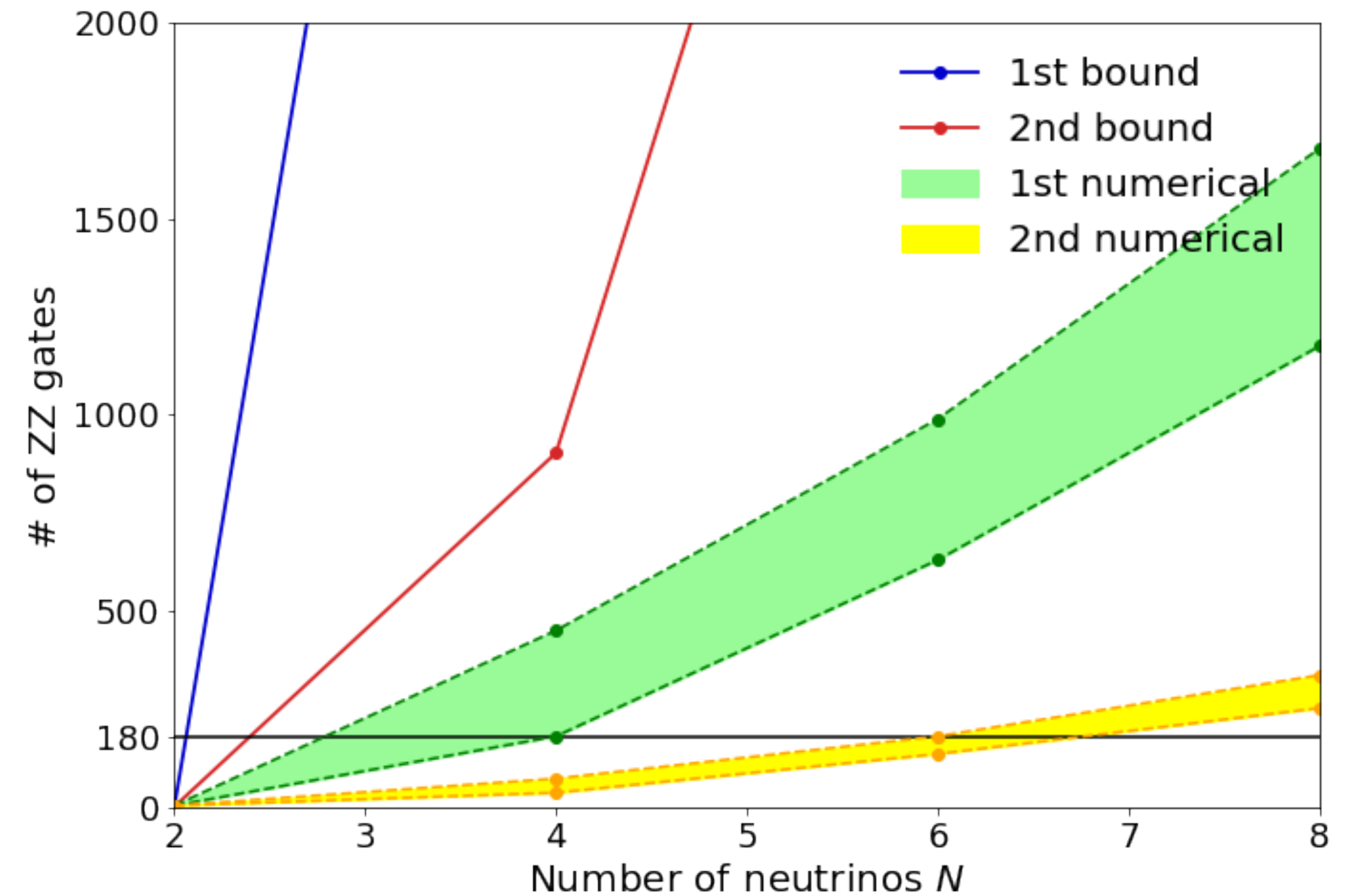


COMPLEXITY SCALING OF THE ALGORITHM

We are interested in systems in which we fix $n_\nu = N/V$ and we look at the scaling with N

Complexity as the number of 2-qubit gates to evolve the system up to T keeping the error $< \epsilon$

- First order Trotter $\mathcal{C}_1 \leq \mathcal{O}\left(\frac{T^2 \mu^2 N^3}{\epsilon}\right)$
- Second order Trotter $\mathcal{C}_2 \leq \mathcal{O}\left(\frac{(T\mu)^{3/2}}{\sqrt{\epsilon}} N^{5/2}\right)$
- Higher order Trotter $\sim N^{2+\delta}$
- Qubitization $\mathcal{C}_Q \leq \mathcal{O}\left(T\mu N^3 + N^2 \log\left(\frac{1}{\epsilon}\right)\right)$



Real cost estimated by calculating the number of steps such that we evolve up to $T = 40\mu^{-1}$ with an error ≤ 0.15

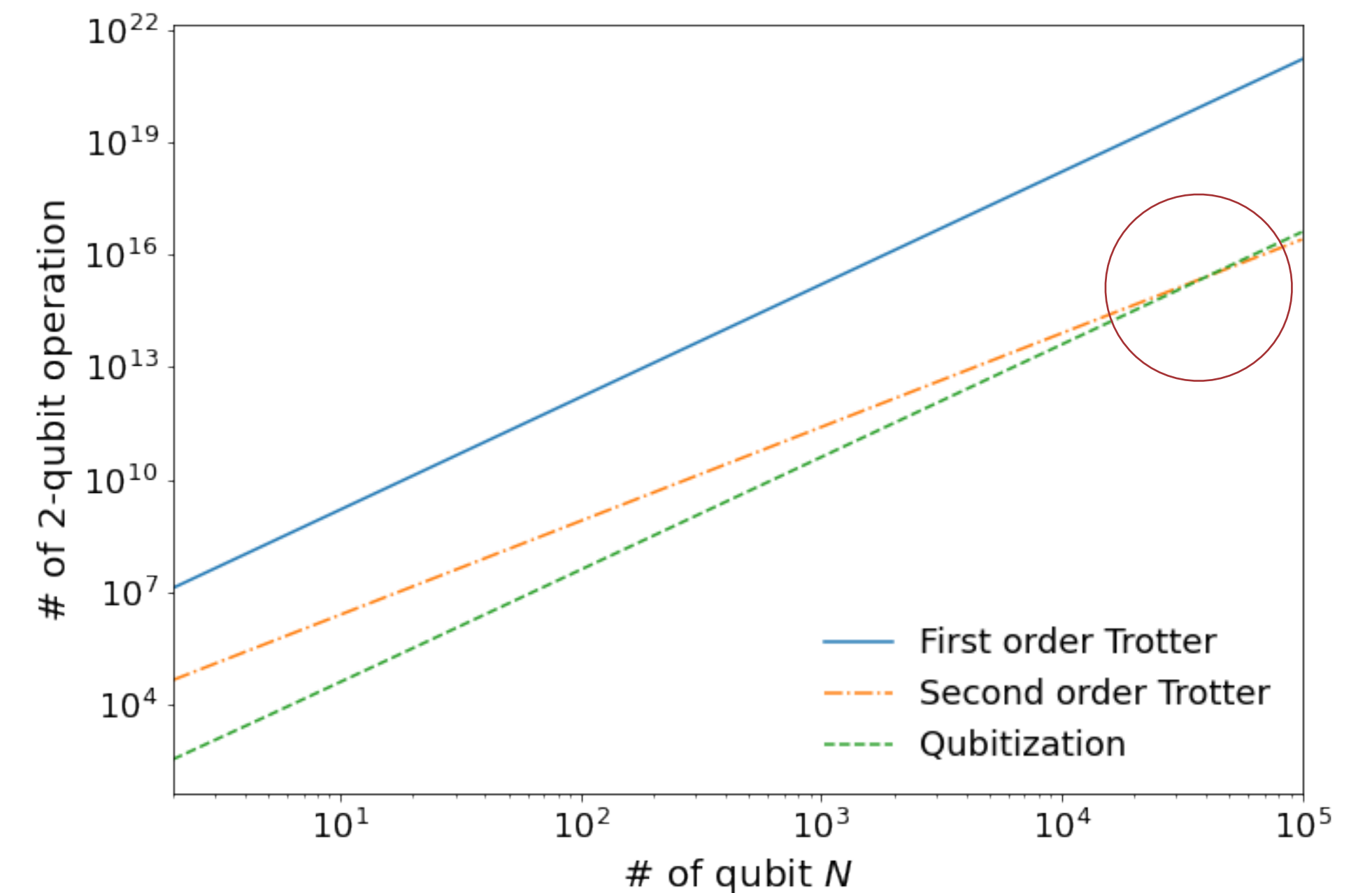
$$\epsilon(dt) = \|U_{approx}(dt) - U_{exact}(dt)\|_\infty$$

$$\epsilon(t) \leq r\epsilon(dt)$$

COMPLEXITY SCALING OF THE ALGORITHM

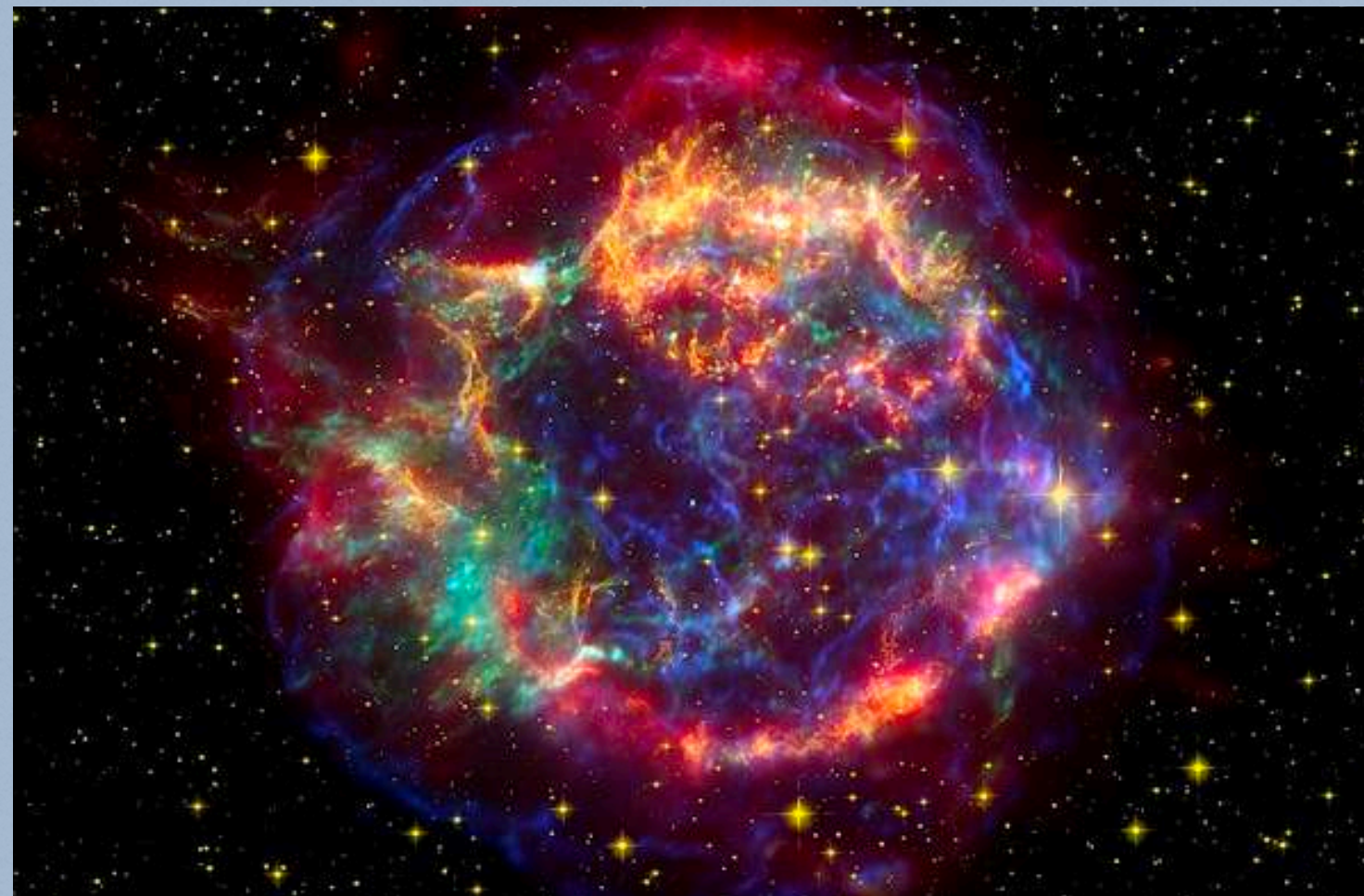
Decomposition type	Single-step error	Number of steps	Circuit complexity
First order Trotter	$\mathcal{O}(dt^2 \mu^2 N)$	$\mathcal{O}\left(\frac{T^2 \mu^2}{\epsilon} N\right)$	$\mathcal{O}\left(\frac{T^2 \mu^2}{\epsilon} N^3\right)$
Second order Trotter	$\mathcal{O}(dt^3 \mu^3 N)$	$\mathcal{O}\left(\frac{T^{3/2} \mu^{3/2}}{\sqrt{\epsilon}} \sqrt{N}\right)$	$\mathcal{O}\left(\frac{T^{3/2} \mu^{3/2}}{\sqrt{\epsilon}} N^{5/2}\right)$
Qubitization	-	$\mathcal{O}(T\mu N + \log(1/\epsilon))$	$\mathcal{O}(T\mu N^3 + \log(1/\epsilon)N^2)$

- Qubitization work well for large time T and small error ϵ
- Trotter method wins for fixed time and error

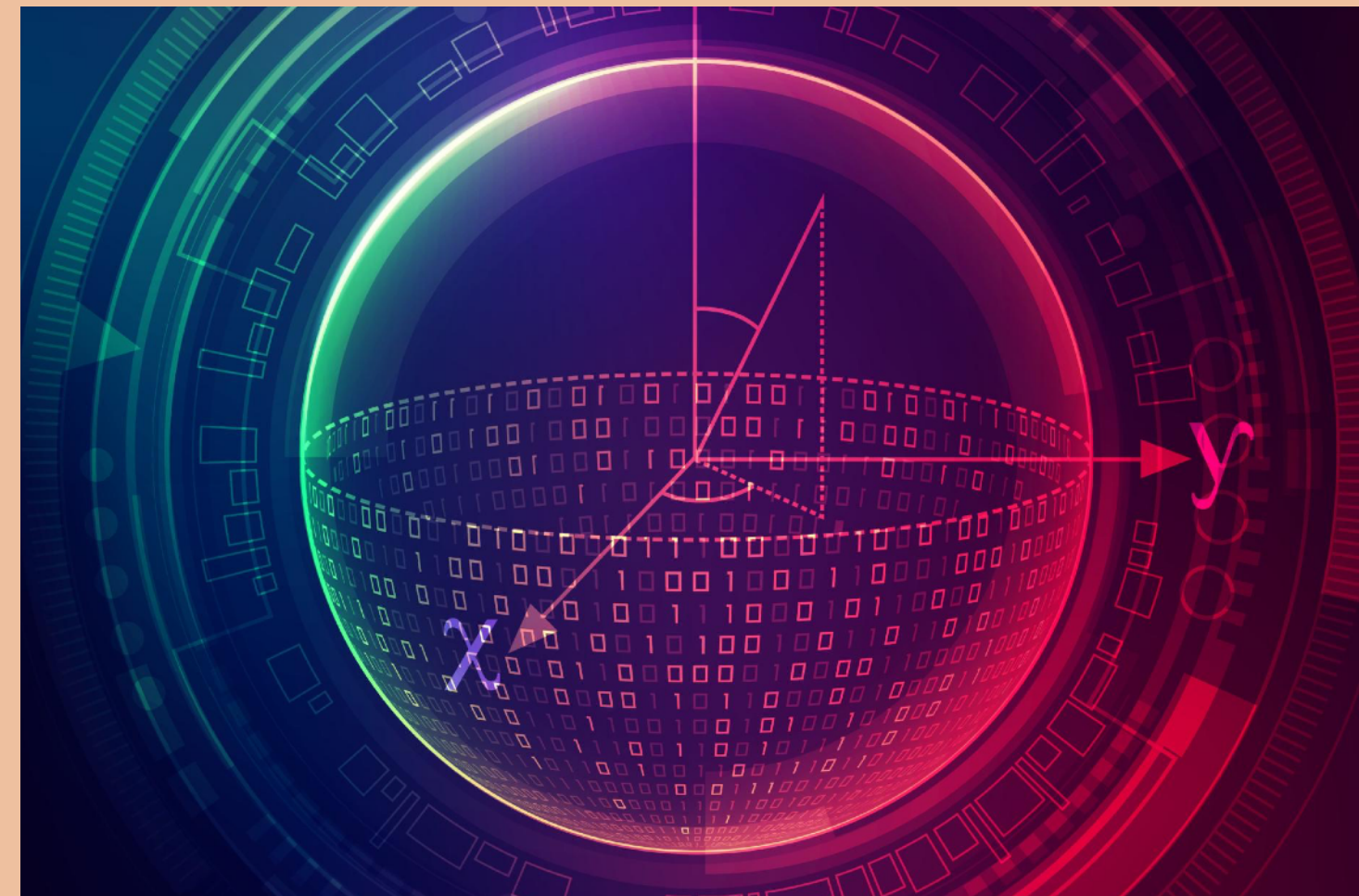


CONCLUSIONS

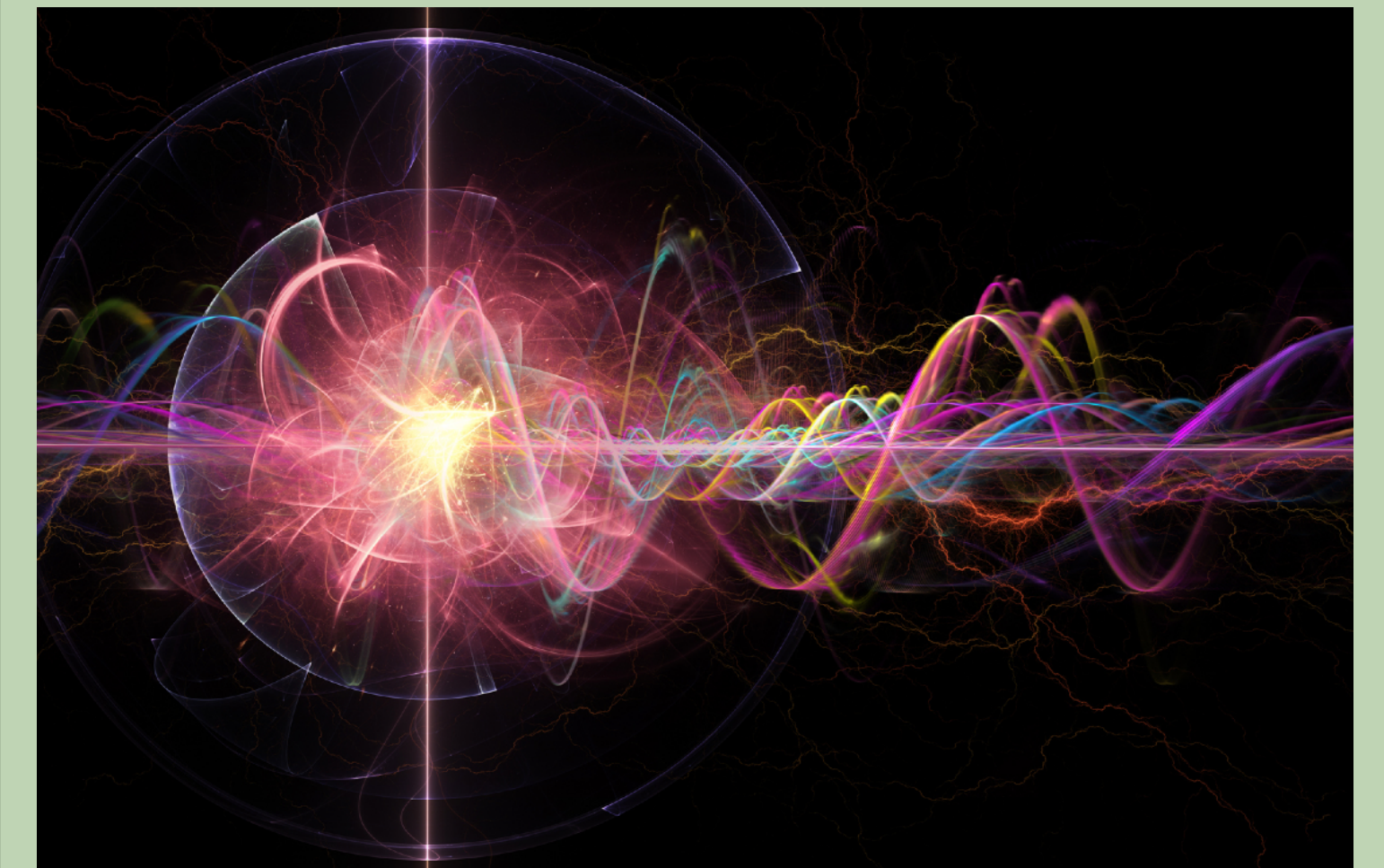
- Flavor dynamics is crucial to describe many effects in core-collapse supernovae
- Collective neutrino oscillations make the problem non linear and interesting to test quantum computing



- QC necessary for full dynamics simulation
- The gate decomposition must be machine aware and circuit optimization is crucial
- Full qubit connectivity allows for more freedom in gate decomposition



- Results are very promising
- We can increase the number of simulated neutrinos
- The complexity of the algorithm scales polynomially with the number of neutrinos





THANK YOU FOR YOUR ATTENTION

Valentina Amitrano
Francesco Pederiva
Alessandro Roggero

**Nuclear and Particle Physics on a Quantum Computer:
Where do we stand now?**

Trento, 6 June 2023 - ECT workshop*

SUPPLEMENTARY MATERIAL

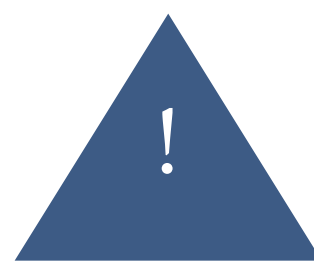
TWO-FLAVOR HAMILTONIAN MODEL

- Vacuum mixing (1-body term)

$$H_{vac} = \Delta \sum_{i=1}^N \vec{b} \cdot \vec{\sigma}_i = \frac{\delta m^2}{4E} \sum_{i=1}^N \left(\sin(2\theta_\nu) X_i - \cos(2\theta_\nu) Z_i \right)$$

- $\nu\nu$ - interaction (2-body term)

$$H_{\nu\nu} = \frac{\mu}{N} \sum_{i<j}^N J_{ij} \vec{\sigma}_i \cdot \vec{\sigma}_j = \frac{\mu}{N} \sum_{i<j}^N J_{ij} \left(X_i \otimes X_j + Y_i \otimes Y_j + Z_i \otimes Z_j \right)$$



$H_{\nu\nu}$ is an all-to-all interaction that makes the problem non-linear

The model:

- $\theta_\nu = 0.195$ mixing angle
- Monochromatic flux $E_i = E \forall i$
- $\vec{b} = \frac{\delta m^2}{4E} (\sin(2\theta_\nu), 0, -\cos(2\theta_\nu))$
- $\Delta = \frac{\delta m^2}{4E}$
- $J_{ij} = 1 - \cos(\theta_{ij})$
- $\theta_{ij} = \arccos(0.9) \frac{|i-j|}{N-1}$
- Energy scale $\mu = \sqrt{2} G_F n_\nu$
- $X_2 = I \otimes I \otimes X \otimes I$
- $X_0 \otimes X_2 = X \otimes I \otimes X \otimes I$

TWO-FLAVOR HAMILTONIAN

- Mass basis $\{\nu_1, \nu_2\}$ and flavor basis $\{\nu_e, \nu_x\}$
- Creation and annihilation operators

$$\begin{pmatrix} a_1^{(\dagger)} \\ a_2^{(\dagger)} \end{pmatrix} = \begin{pmatrix} \cos(\theta_\nu) & -\sin(\theta_\nu) \\ \sin(\theta_\nu) & \cos(\theta_\nu) \end{pmatrix} \begin{pmatrix} a_e^{(\dagger)} \\ a_x^{(\dagger)} \end{pmatrix}$$

- On the mass basis:

$$H_{vac} = E_1 a_1^\dagger a_1 + E_2 a_2^\dagger a_2$$

- On the flavor one:

$$H_{vac} = \frac{\delta m^2}{4E} \sin(2\theta_\nu)(a_e^\dagger a_x + a_x^\dagger a_e) + \frac{\delta m^2}{4E} \cos(2\theta_\nu)(a_x^\dagger a_x - a_e^\dagger a_e)$$

- Mapping:

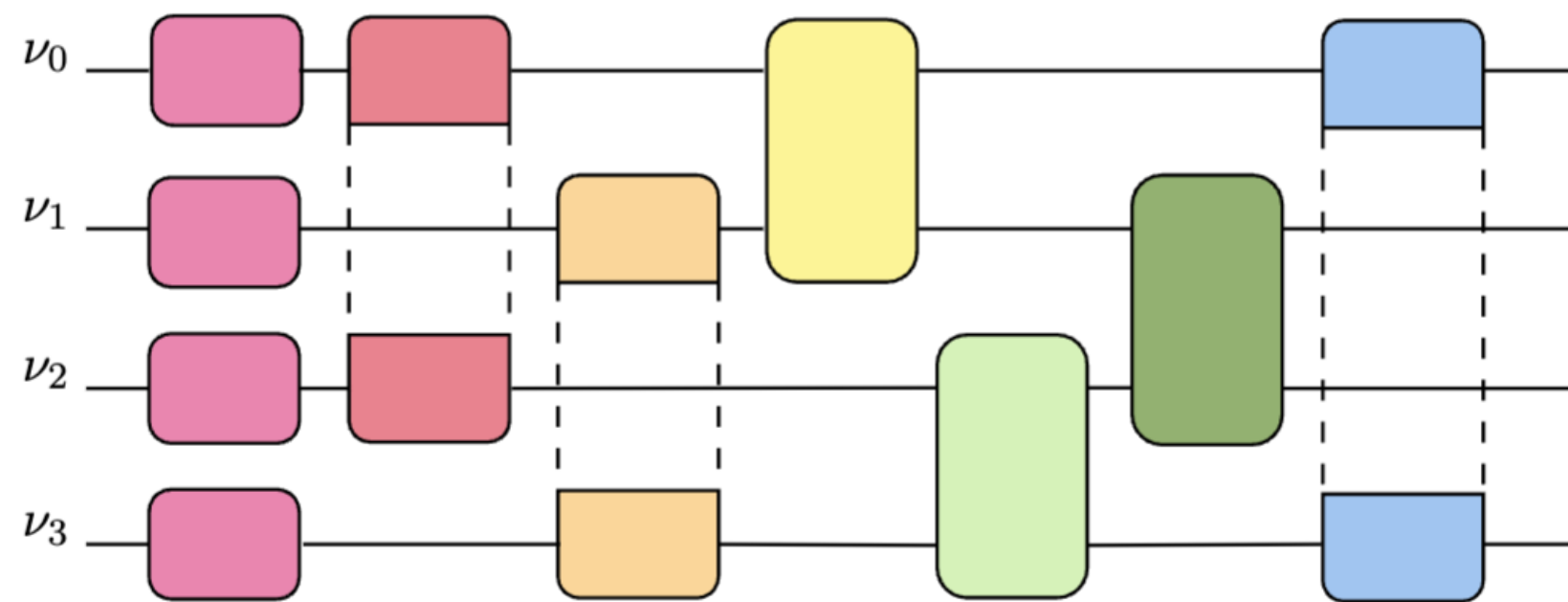
$$\sigma_z = a_e^\dagger a_e - a_x^\dagger a_x \text{ and } \sigma_x = a_e^\dagger a_x + a_x^\dagger a_e$$

- We have

$$H_{vac} = \frac{\delta m^2}{4E} (\sin(2\theta_\nu)X - \cos(2\theta_\nu)Z)$$

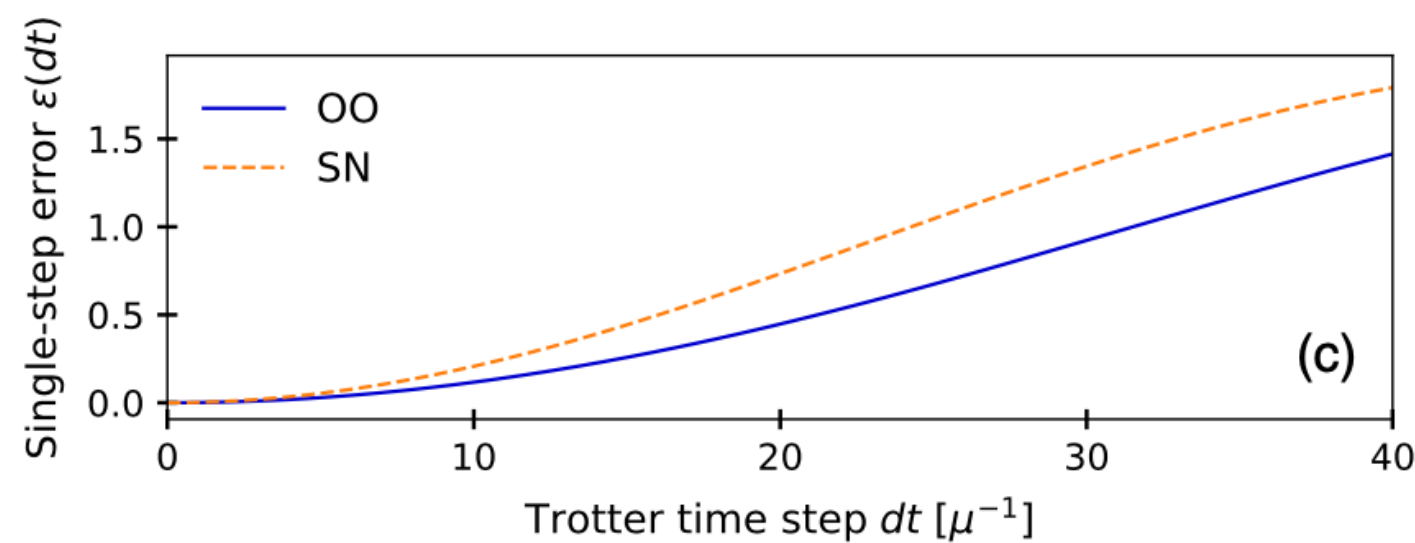
QUBIT CONNECTIVITY: TROTTER ERROR

Optimal Circuit for full connectivity

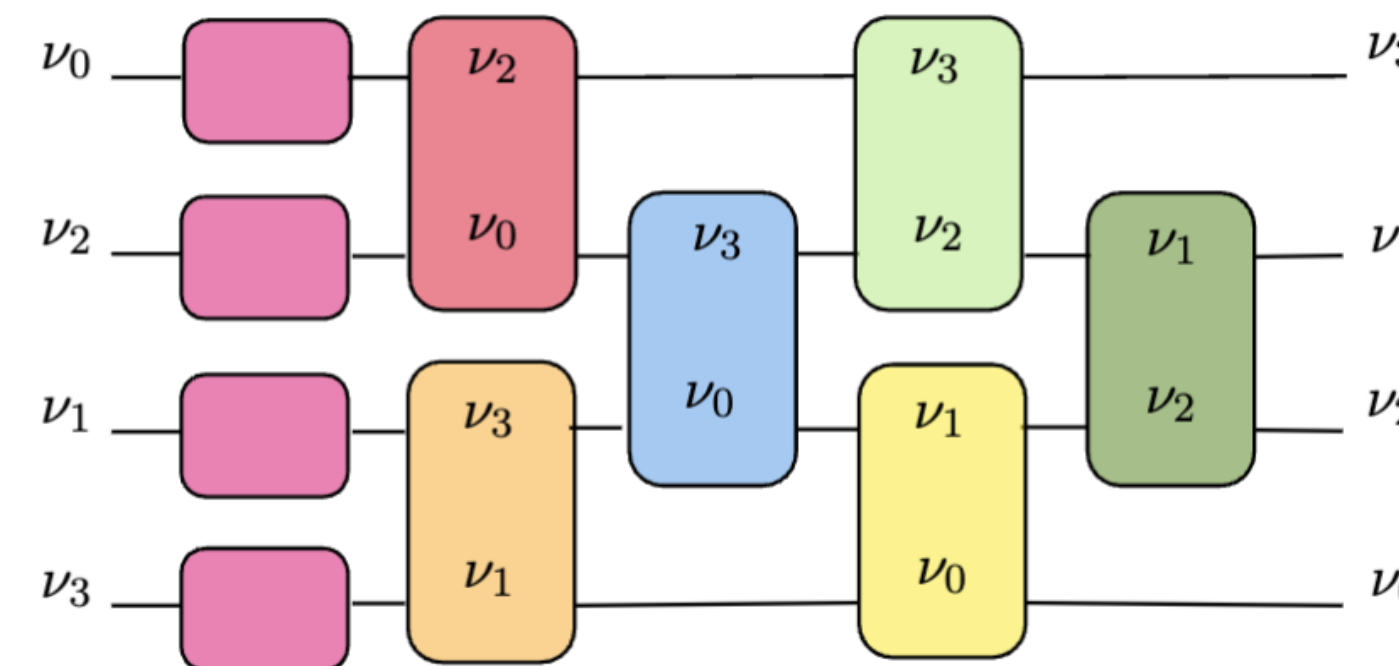


Full freedom in the pair ordering

*V. Amitrano et al.
Phys. Rev. D 107,
023007 (2023)*

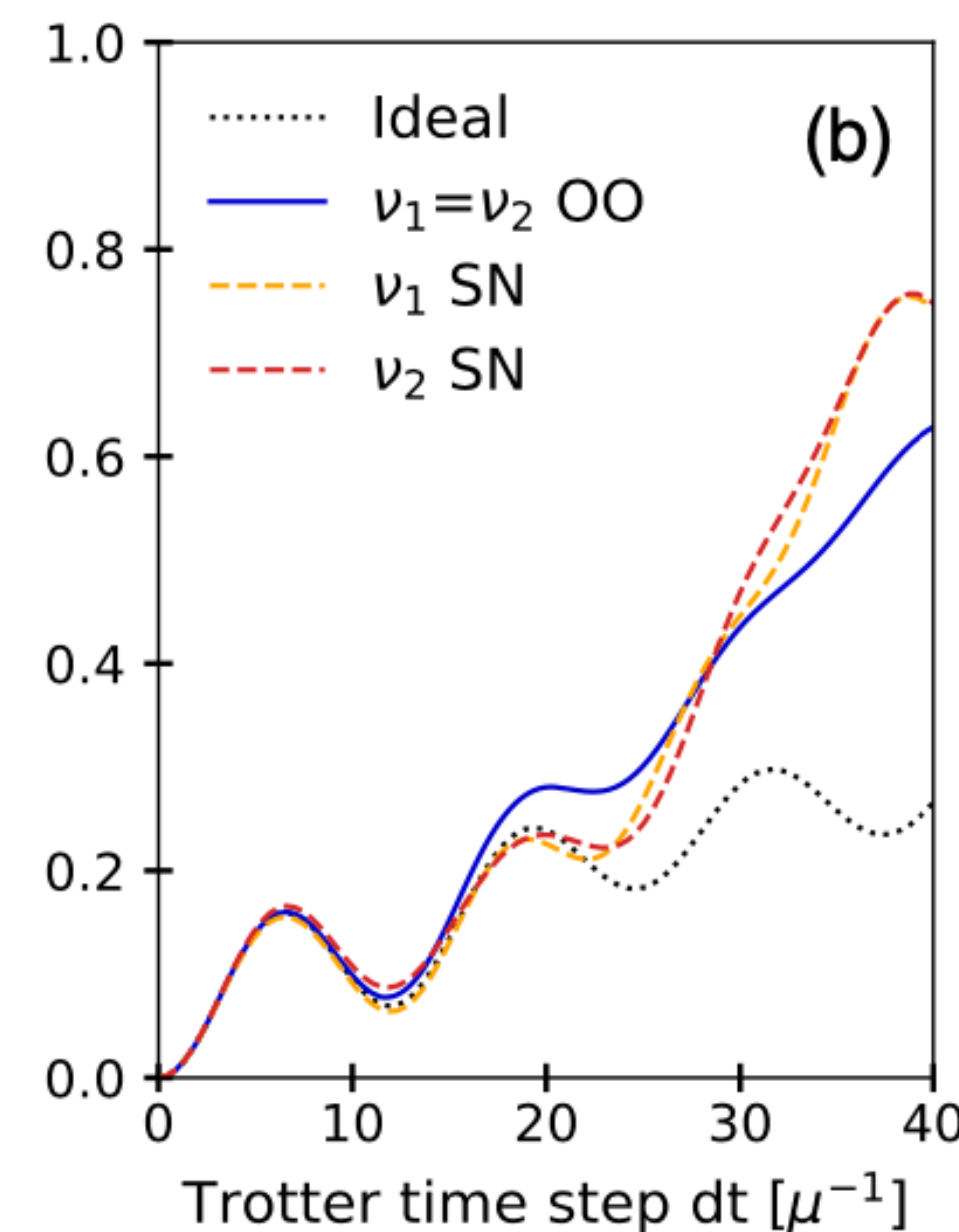
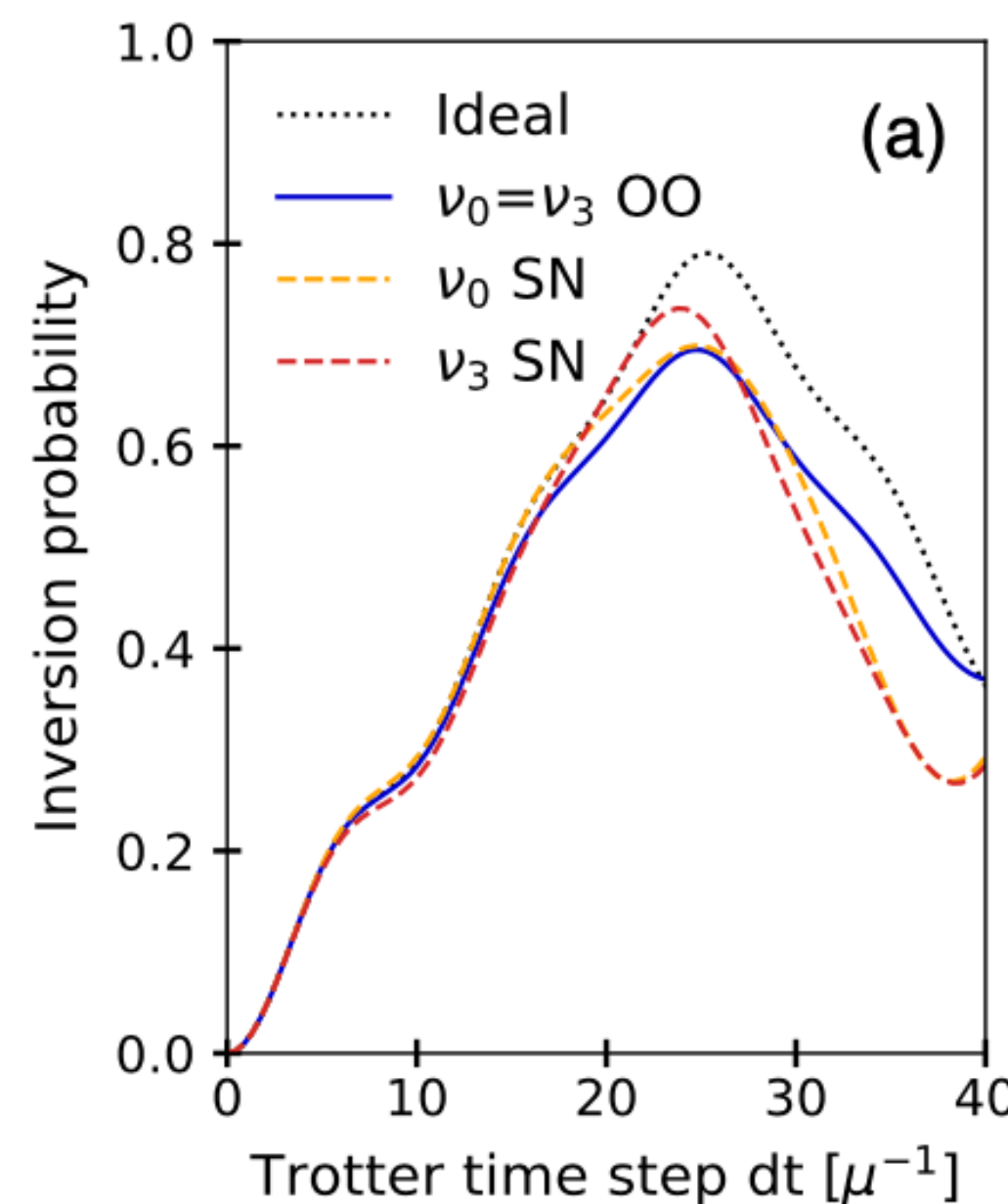


Swap network for linear connectivity



*Hall, A. Roggero
et. al (2021)*

Full freedom in the initial encoding, but the pair ordering is fixed. Any order can be obtained at the price of adding additional SWAP gates.

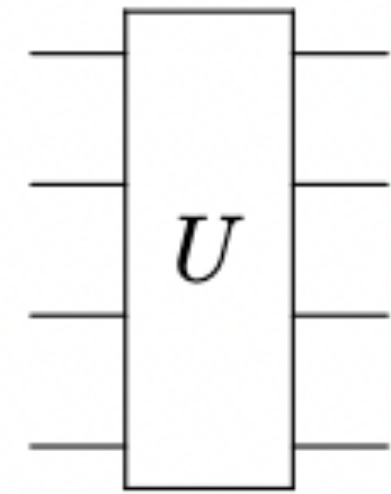


SINGLE VS MULTIPLE EVOLUTION STEPS

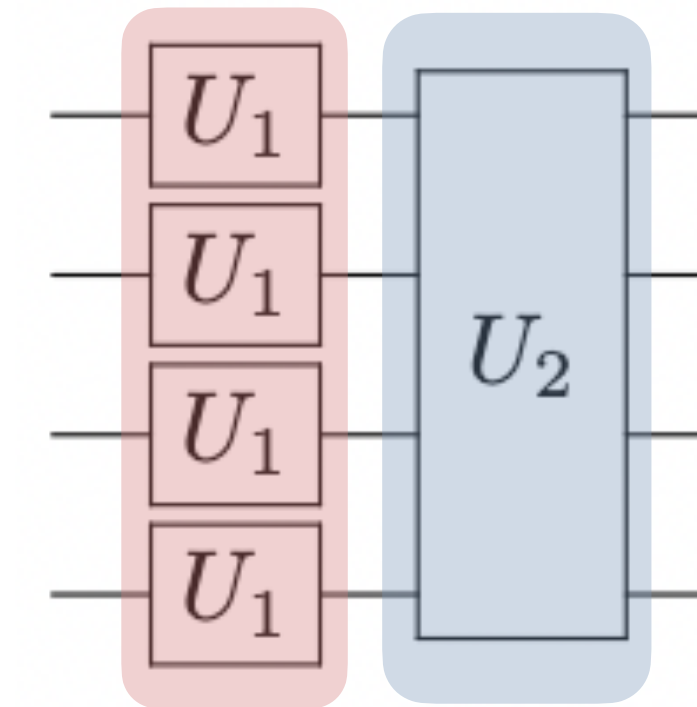
Single Trotter
step

$$U(dt) = e^{-iHdt}$$

$$H = H_{vac} + H_{\nu\nu}$$

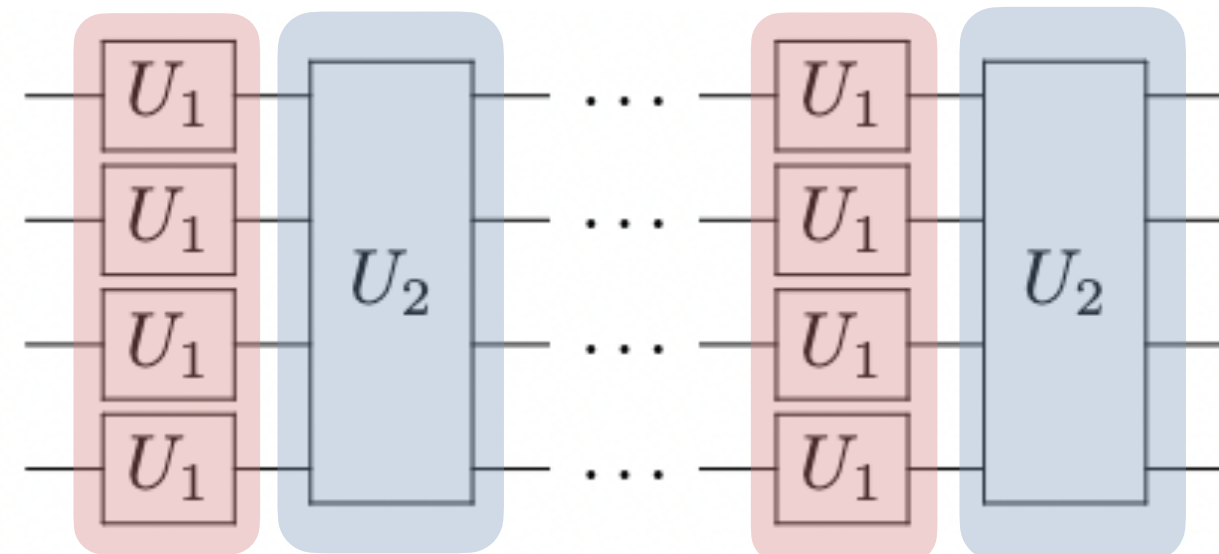


$$U(dt) = U_2(dt)U_1(dt)$$

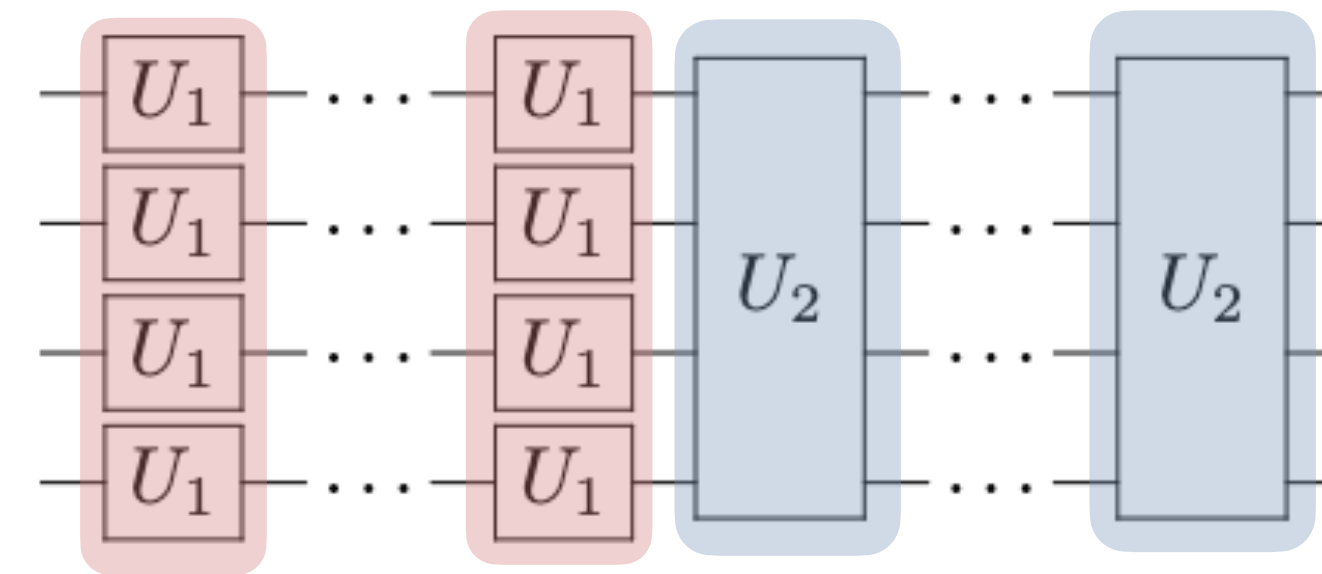


Evolution
using
multiple
Trotter steps

$$U(T) = \prod_{k=1}^r U_2\left(\frac{T}{r}\right) U_1\left(\frac{T}{r}\right)$$



$$U(T) = \prod_{k=1}^r U_2\left(\frac{T}{r}\right) \prod_{k=1}^r U_1\left(\frac{T}{r}\right)$$

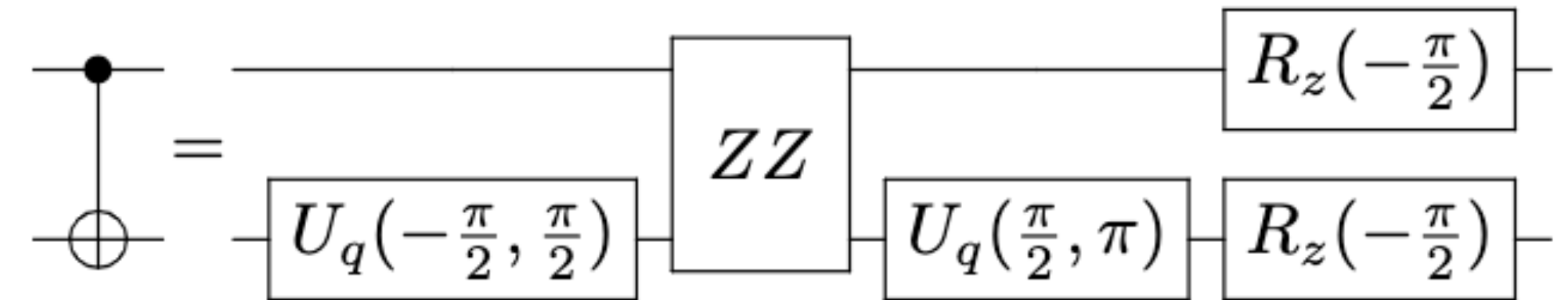


Quantum native gate set

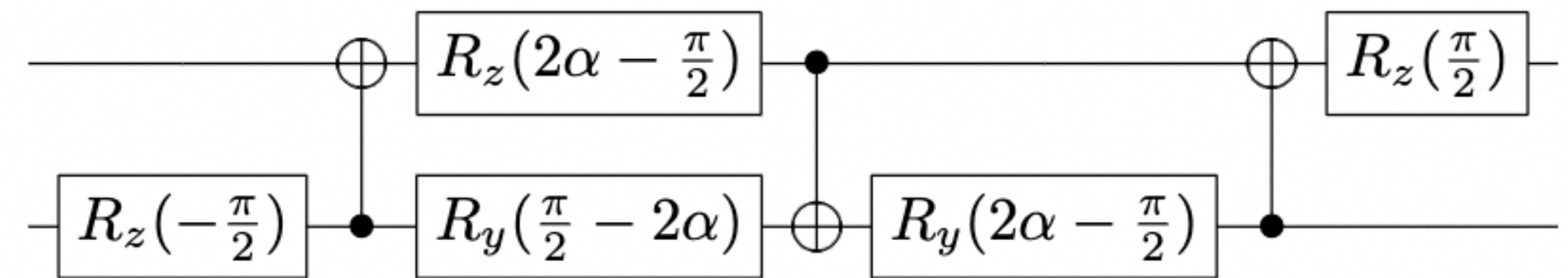
- $R_z(\lambda) = \begin{pmatrix} e^{-i\lambda/2} & 0 \\ 0 & e^{i\lambda/2} \end{pmatrix}$
- $U_q(\theta, \varphi) = \begin{pmatrix} \cos \theta/2 & -ie^{-i\varphi} \sin \theta/2 \\ -ie^{i\varphi} \sin \theta/2 & \cos \theta/2 \end{pmatrix}$

- $ZZ = e^{-i\frac{\pi}{4}} Z \otimes Z = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & i & 0 & 0 \\ 0 & 0 & i & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

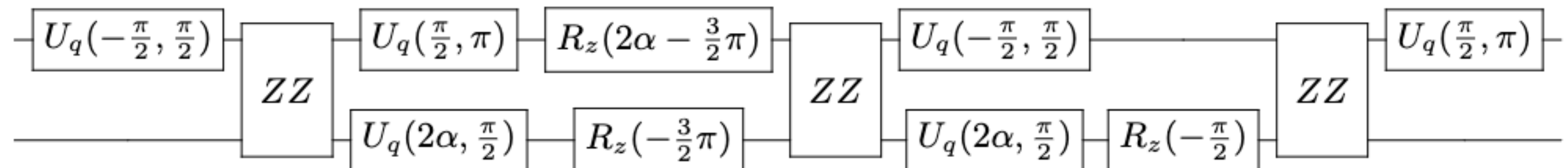
ZZ-based CNOT gate



CNOT-based u_{ij}



ZZ-based u_{ij}



TROTTER ERROR SCALING

First order

$$U_2(dt) \approx \mathcal{L}_1(dt) = \prod_{K=1}^{\Gamma} e^{-ih_{ij}dt}$$

$$\varepsilon(dt) \leq \frac{dt^2}{2} \sum_{K=1}^{\Gamma} \left\| \sum_{L=K+1}^{\Gamma} [h_K, h_L] \right\|$$

$$\varepsilon(dt) \leq 12dt^2\mu^2 \frac{\Theta^2}{N^2} \binom{N}{3} = \mathcal{O}(dt^2\mu^2N)$$

$$\varepsilon(T) \leq r\varepsilon(dt) \quad T = rdt$$

$$r \leq 12 \frac{T^2\mu^2\Theta^2}{\epsilon N^2} \binom{N}{3} = \mathcal{O}\left(\frac{T^2\mu^2N}{\epsilon}\right)$$

$$\mathcal{C} \leq \binom{N}{2} r$$

$$\mathcal{C}_1 = \mathcal{O}\left(\frac{T^2\mu^2N^3}{\epsilon}\right)$$

TROTTER ERROR SCALING

Second order

$$U(dt) \approx \mathcal{L}_2(dt) = \mathcal{L}_1\left(\frac{dt}{2}\right) \mathcal{L}_1^\dagger\left(-\frac{dt}{2}\right)$$

$$\begin{aligned} \varepsilon(dt) \leq & \frac{dt^3}{12} \sum_K^\Gamma \left\| \sum_{L>K}^\Gamma \sum_{M>K}^\Gamma [h_L, [h_M, h_K]] \right\| + \\ & + \frac{dt^3}{24} \sum_K^\Gamma \left\| \sum_{L>K}^\Gamma [h_K, [h_K, h_L]] \right\| \end{aligned}$$

$$\varepsilon(dt) \leq dt^3 \frac{\mu^3 \Theta^3}{N^3} \left[20 \binom{N}{3} + 56 \binom{N}{4} \right] = \mathcal{O}\left(dt^3 \mu^3 N\right)$$

$$\varepsilon(T) \leq r \varepsilon(dt) \quad T = rdt$$

$$r \leq \frac{(T\mu\Theta)^{3/2}}{\sqrt{\epsilon} N^{3/2}} \sqrt{20 \binom{N}{3} + 56 \binom{N}{4}} = \mathcal{O}\left(\frac{T^{3/2} \mu^{3/2} \sqrt{N}}{\sqrt{\epsilon}}\right)$$

$$\mathcal{C} \leq \left(2 \binom{N}{2} - \frac{N}{2} \right) r$$

$$\mathcal{C}_2 \leq \left(2 \binom{N}{2} - \frac{N}{2} \right) r = \mathcal{O}\left(\frac{(T\mu)^{3/2}}{\sqrt{\epsilon}} N^{5/2}\right)$$

- Number of repetitions $M = 200$
- Bayesian approach
 - Probability distribution of obtaining m times the output $|q\rangle$:

$$\mathcal{P}_b(m|p) = \binom{M}{m} p^m (1-p)^{M-m}$$

Likelihood distribution:
Binomial distribution

- Bayes theorem:

$$\mathcal{P}(p|m) = \frac{\mathcal{P}(m|p)\mathcal{P}(p)}{\mathcal{P}(m)}$$

Posterior distribution

A priori distribution: Beta distribution

$$\mathcal{B}(\alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} p^{\alpha-1} (1-p)^{\beta-1}$$

- Prior conjugate

$$\mathcal{B}(\alpha', \beta') = \frac{\mathcal{P}_b(m|p)\mathcal{B}(\alpha, \beta)}{\int dq \mathcal{P}_b(m|p)\mathcal{B}(\alpha, \beta)} \text{ where } \alpha' = \alpha + m \text{ and } \beta' = \beta + M - m$$

- $\alpha = 1$ and $\beta = 1$. We used $\mathcal{B}(\alpha', \beta')$ as posterior distribution and look for:
 - $\mathcal{P}(p_{min} < p < p_{max}) = 0.68$