



Digital quantum simulation of molecular dynamics and control

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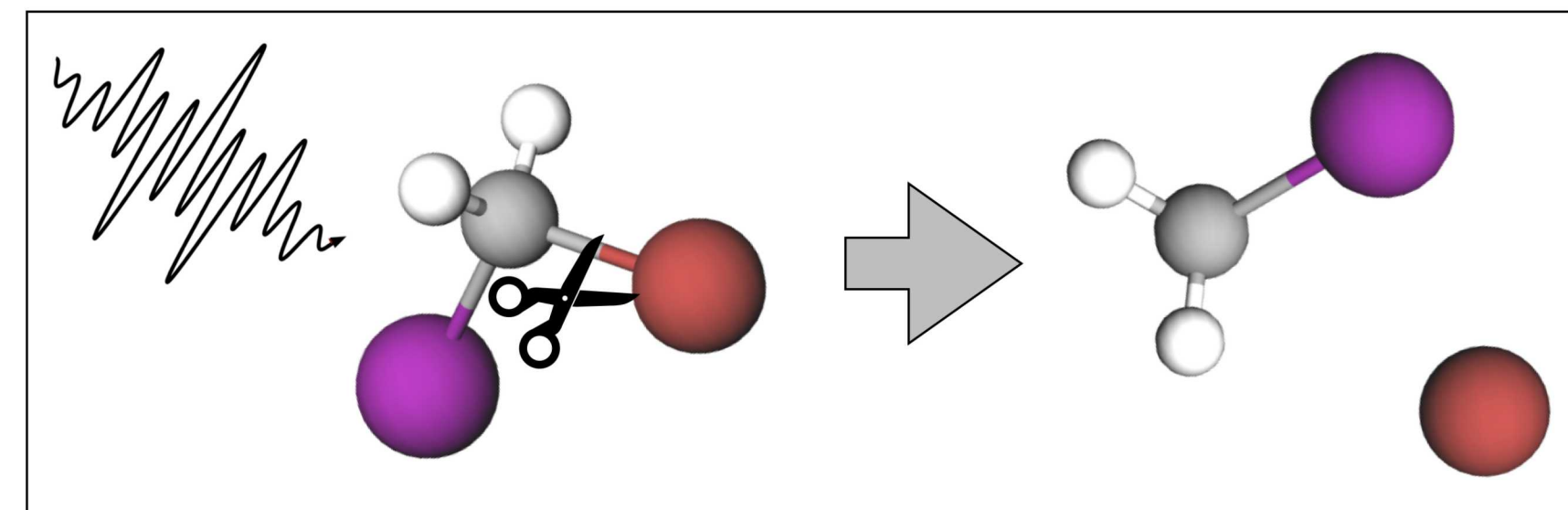
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Quantum optimal control



Aim is to design a time-dependent field $f(t, \{\theta_i\})$, $t \in [0, T]$, to steer a quantum system towards a desired control target.
This is posed as the search for $\min_{\{\theta_i\}} J[T, \{\theta_i\}]$.

Simulations have two components:

1. Evaluation of control objective functional $J[T, \{\theta_i\}]$
2. Update of control parameters $\{\theta_i\}$

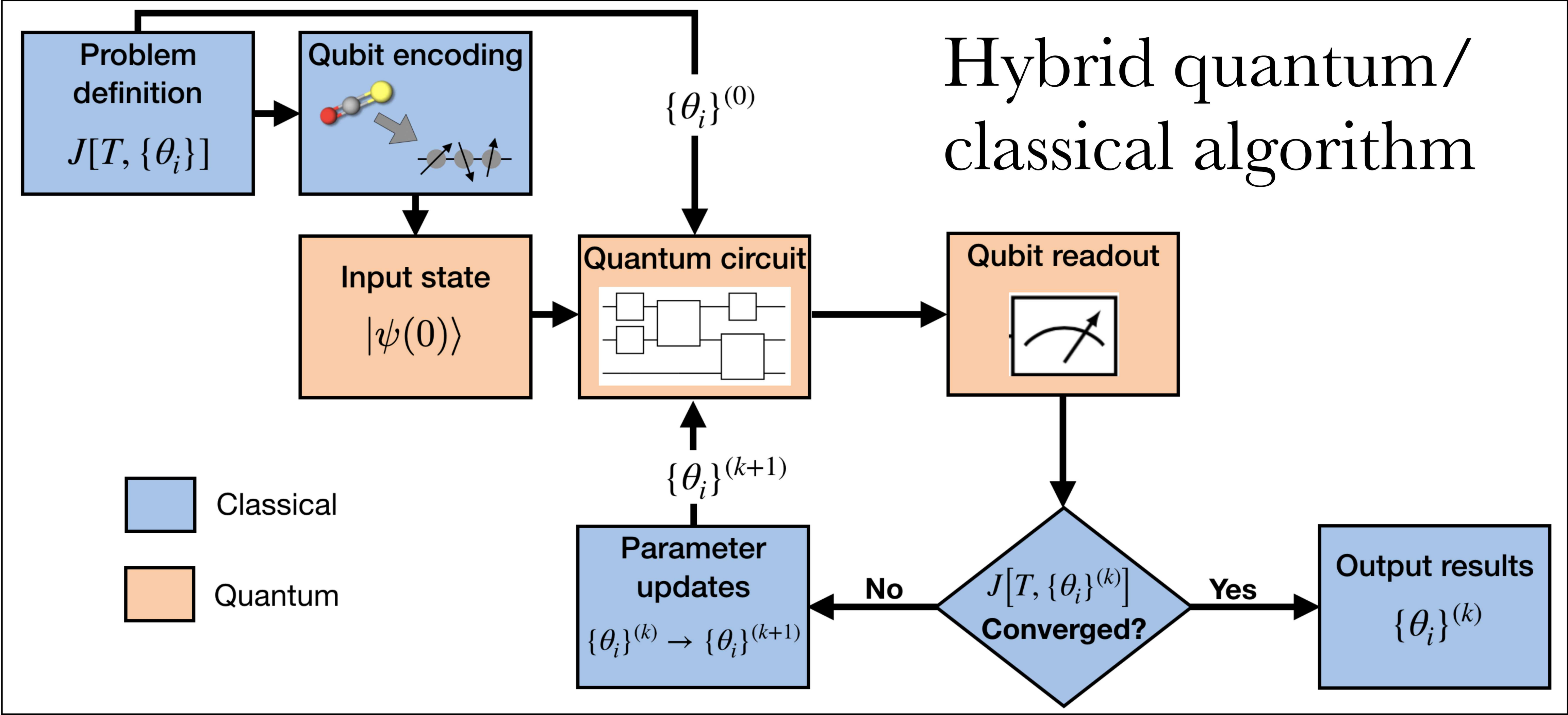
How well is desired control target reached?

Involves simulating the dynamics of a quantum system under the influence of an applied field

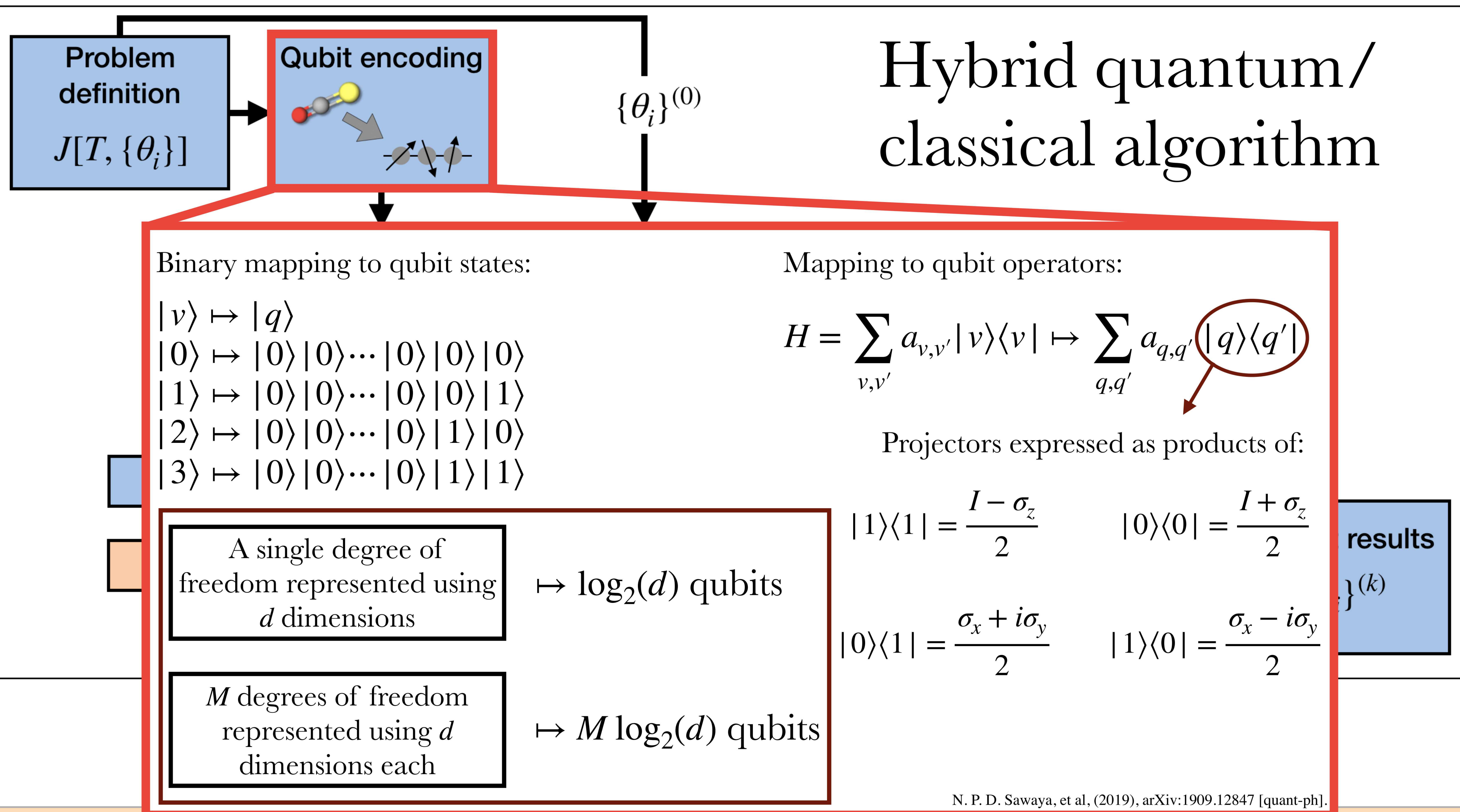
Iteratively update $\{\theta_i\}$ until $J[T, \{\theta_i\}]$ converges

- Computationally demanding
- Computer memory/time costs scale exponentially in # quantum degrees of freedom

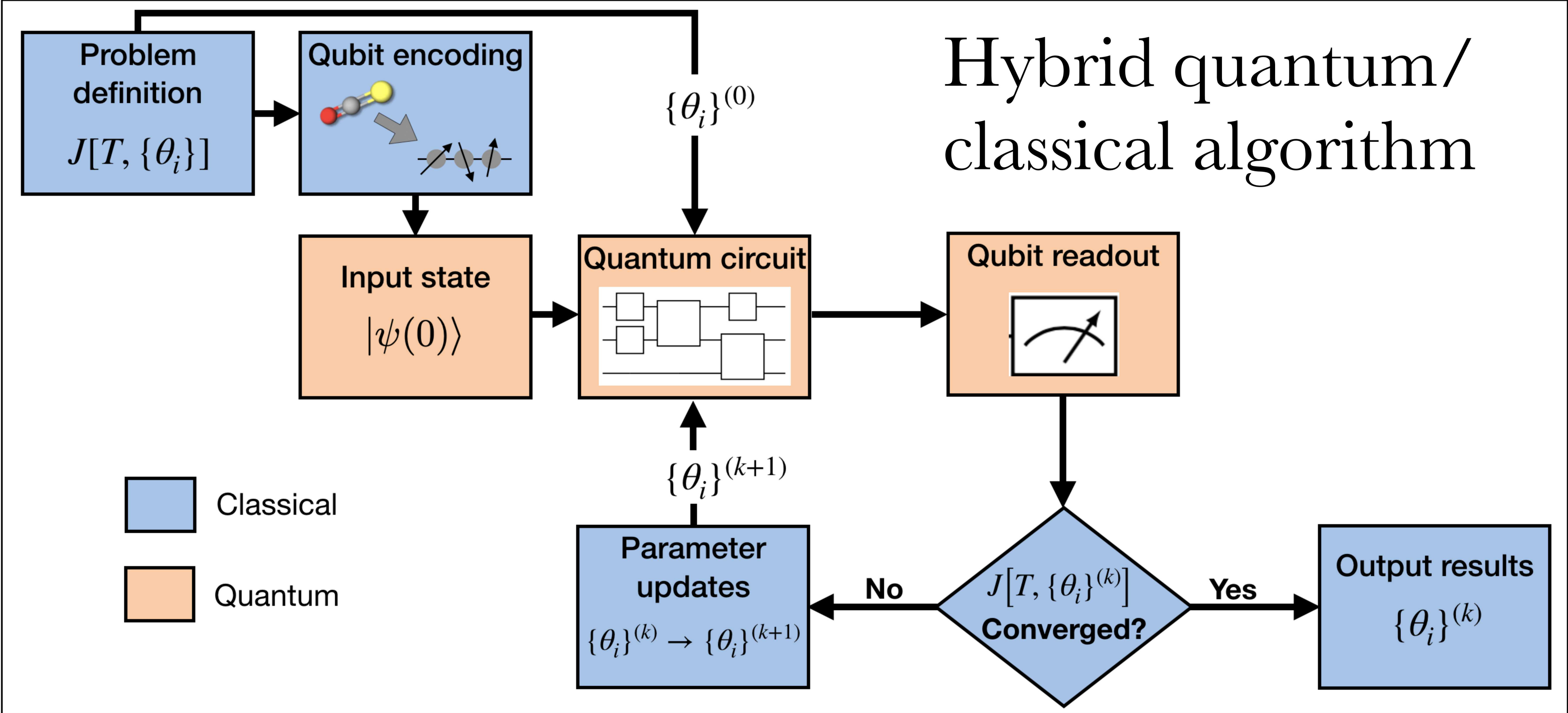
...not on a quantum computer

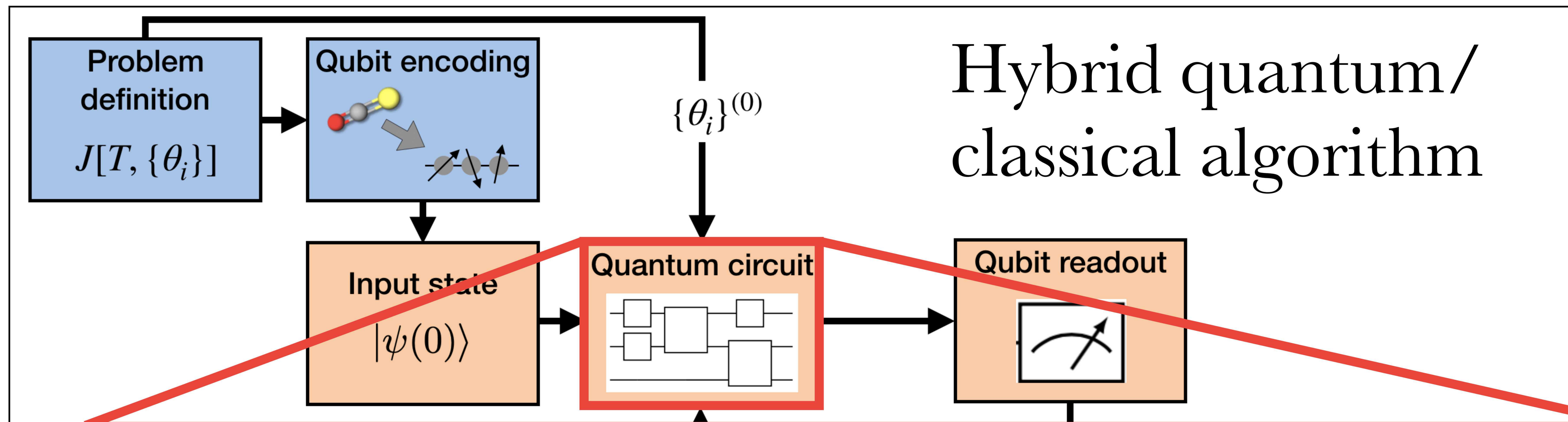


Hybrid quantum/ classical algorithm



N. P. D. Sawaya, et al, (2019), arXiv:1909.12847 [quant-ph].





Time evolution of initial state $|\psi(0)\rangle$ to terminal state $|\psi(T)\rangle$ simulated as

$$|\psi(T)\rangle = U(T,0) |\psi(0)\rangle$$

Approximate $U(T,0)$ as time-ordered product,

$$U(T,0) = U(T, T - \Delta t) \cdots U(2\Delta t, \Delta t) U(\Delta t, 0)$$

Time evolution over each Δt then approximated using product formulas,

$$\begin{aligned} U(t + \Delta t, t) &\approx U_{PF1}(t + \Delta t, t) \\ &= \left(e^{-iH_1(t)\Delta t/n} e^{-iH_2(t)\Delta t/n} \cdots e^{-iH_L(t)\Delta t/n} \right)^n \end{aligned}$$

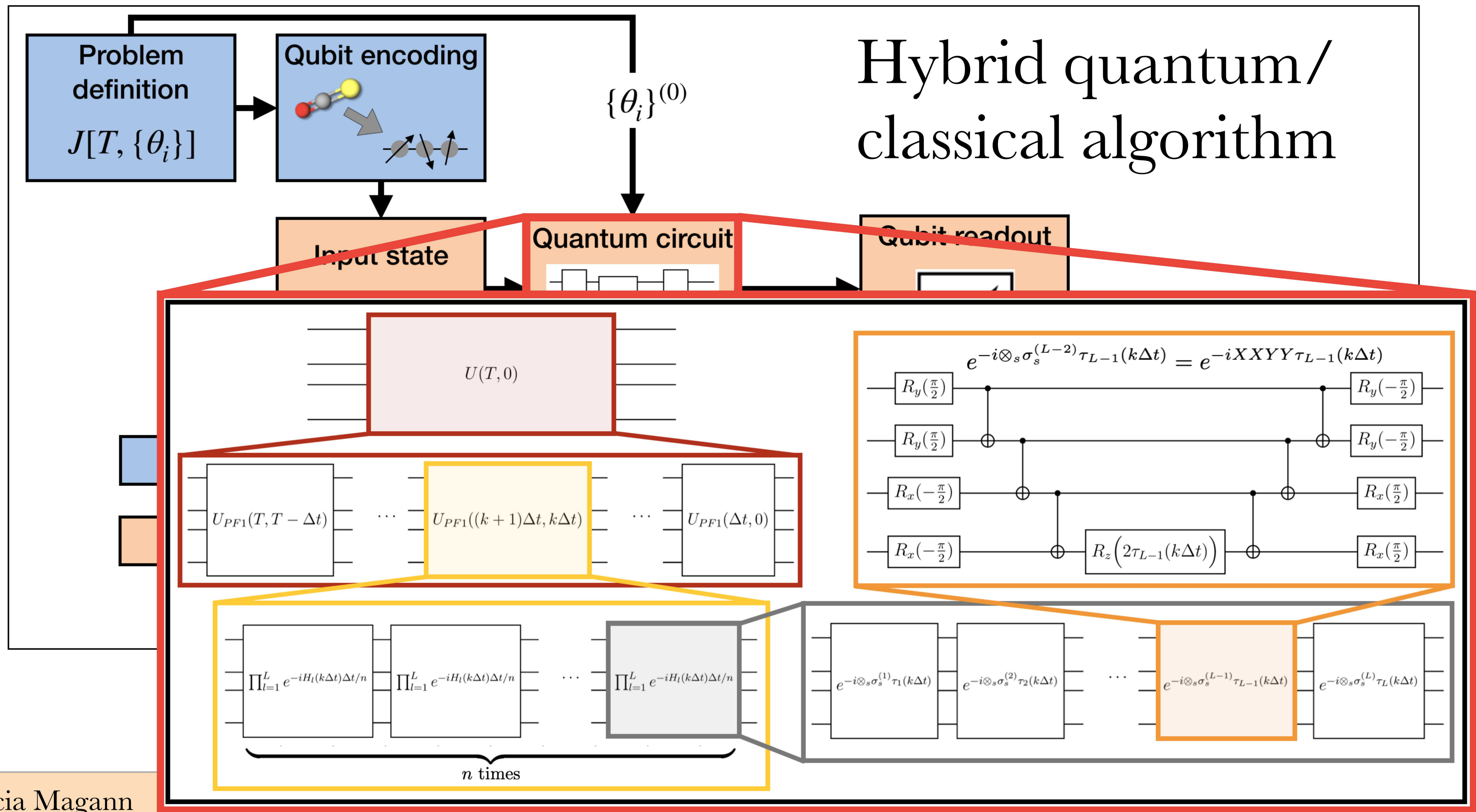
Errors can be reduced with higher order product formulas,

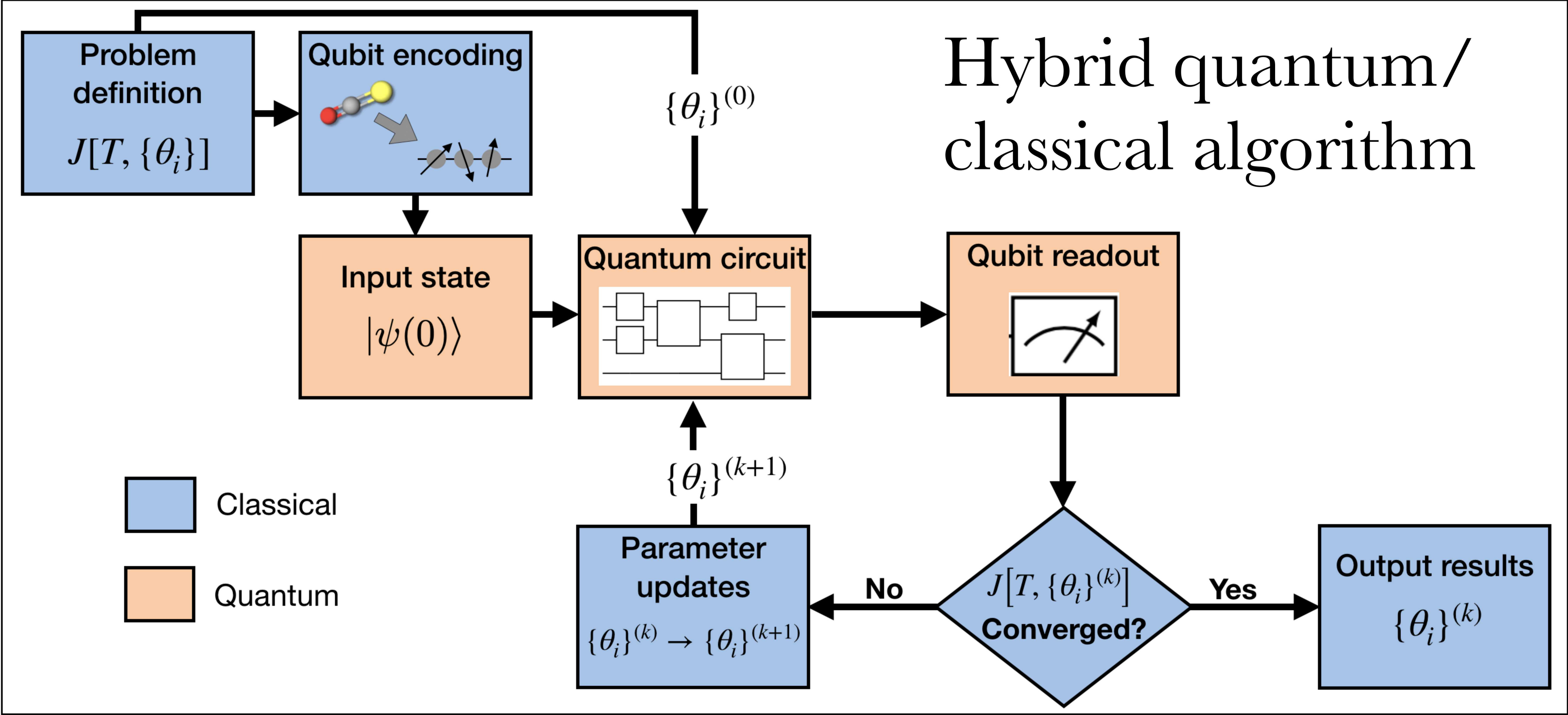
$$U_{PF(2p)}(t + \Delta t, t) = \left(S_{2p}(t, \frac{-i\Delta t}{n}) \right)^n$$

defined recursively as

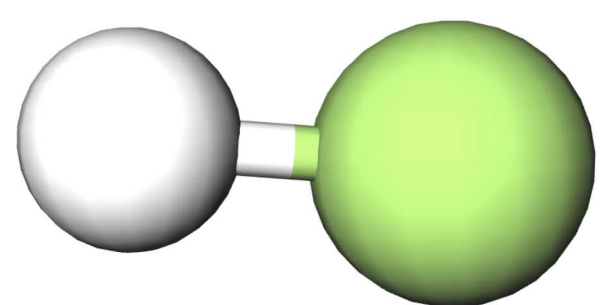
$$\begin{aligned} S_{2p}(t, \lambda) &= \left(S_{2p-2}(t, \gamma_p \lambda) \right)^2 S_{2p-2}(t, (1 - 4\gamma_p)\lambda) \left(S_{2p-2}(\gamma_p \lambda) \right)^2 \\ S_2(t, \lambda) &= \left(\prod_{l=1}^L e^{H_l(t)\lambda/2} \right) \left(\prod_{l=L}^1 e^{H_l(t)\lambda/2} \right) \quad \gamma_p = \frac{1}{4 - 4^{1/(2p-1)}} \end{aligned}$$

Hybrid quantum/ classical algorithm





Numerical illustrations



Controlled bond stretching in HF

Goal: drive HF to target stretch $\gamma = 1.5r_0$ at terminal time T

$$J[T, \{\theta_i\}] = (\langle \psi(T) | r | \psi(T) \rangle - \gamma)^2$$

The field-induced vibrational dynamics of HF are simulated by modeling it as an anharmonic Morse oscillator

$$H(t) = \frac{p^2}{2m} + V(r) - \mu(r)f(t, \{\theta_i\})$$

$$V(r) = D(1 - e^{-\alpha(r-r_0)})^2 - D$$

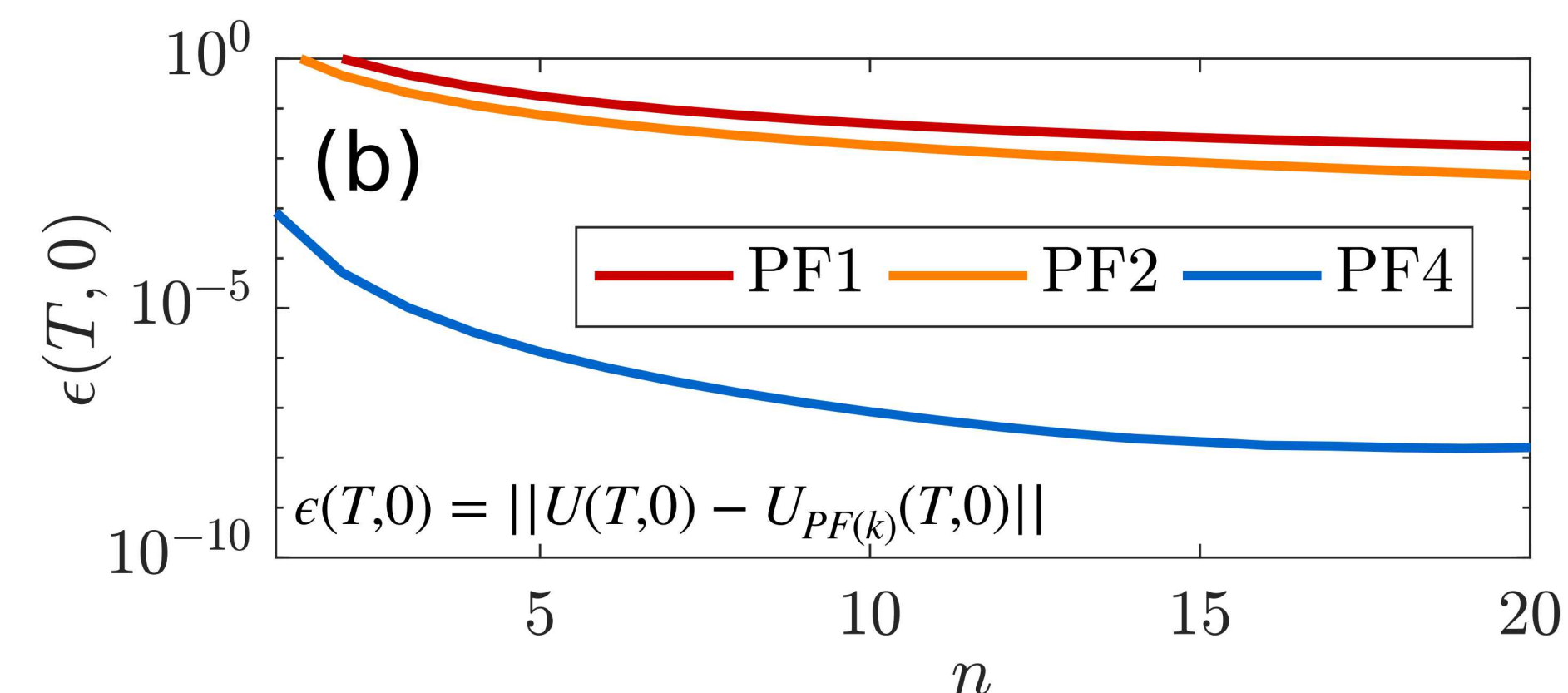
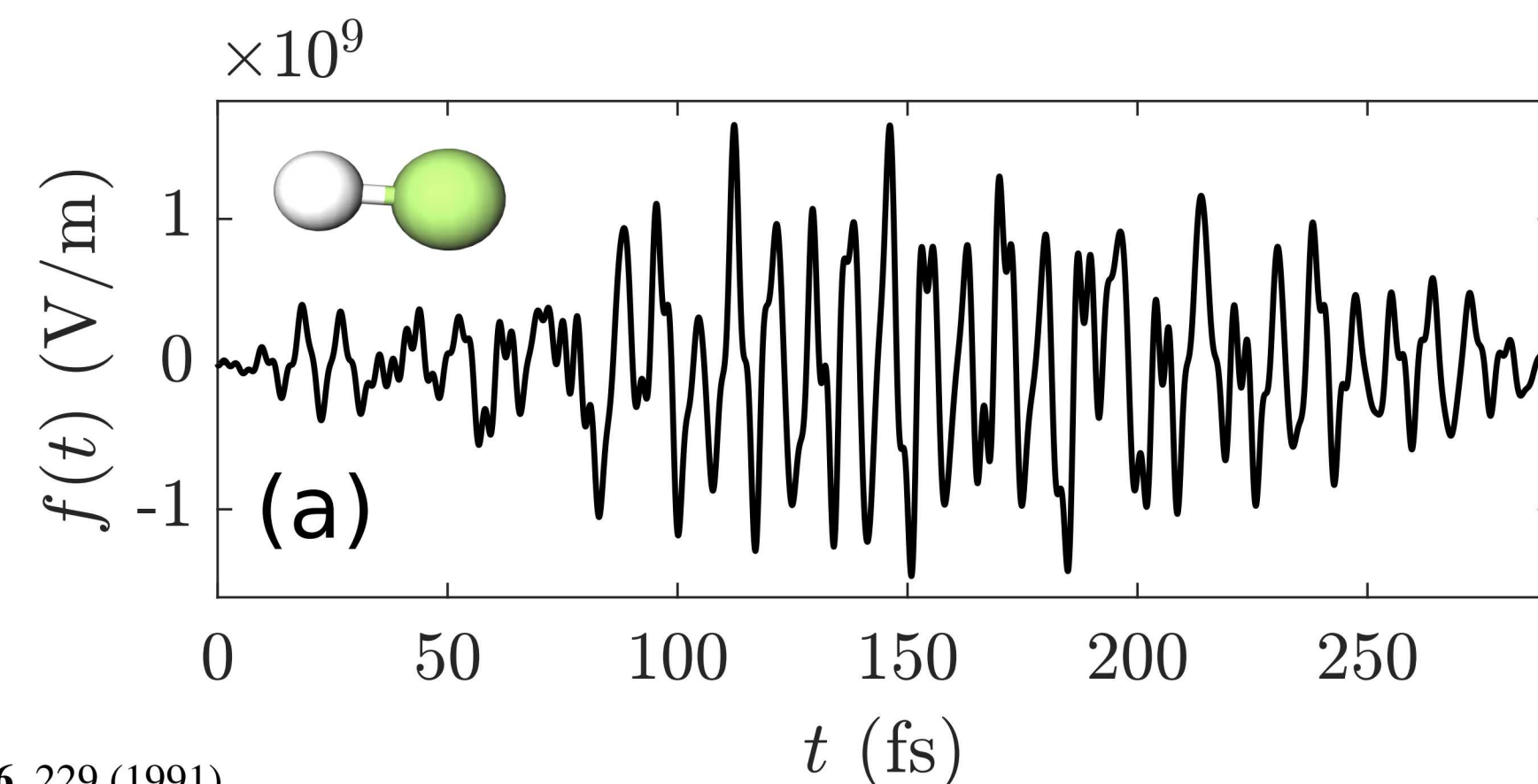
$$\mu(r) = \mu_e r e^{-\beta r^4}$$

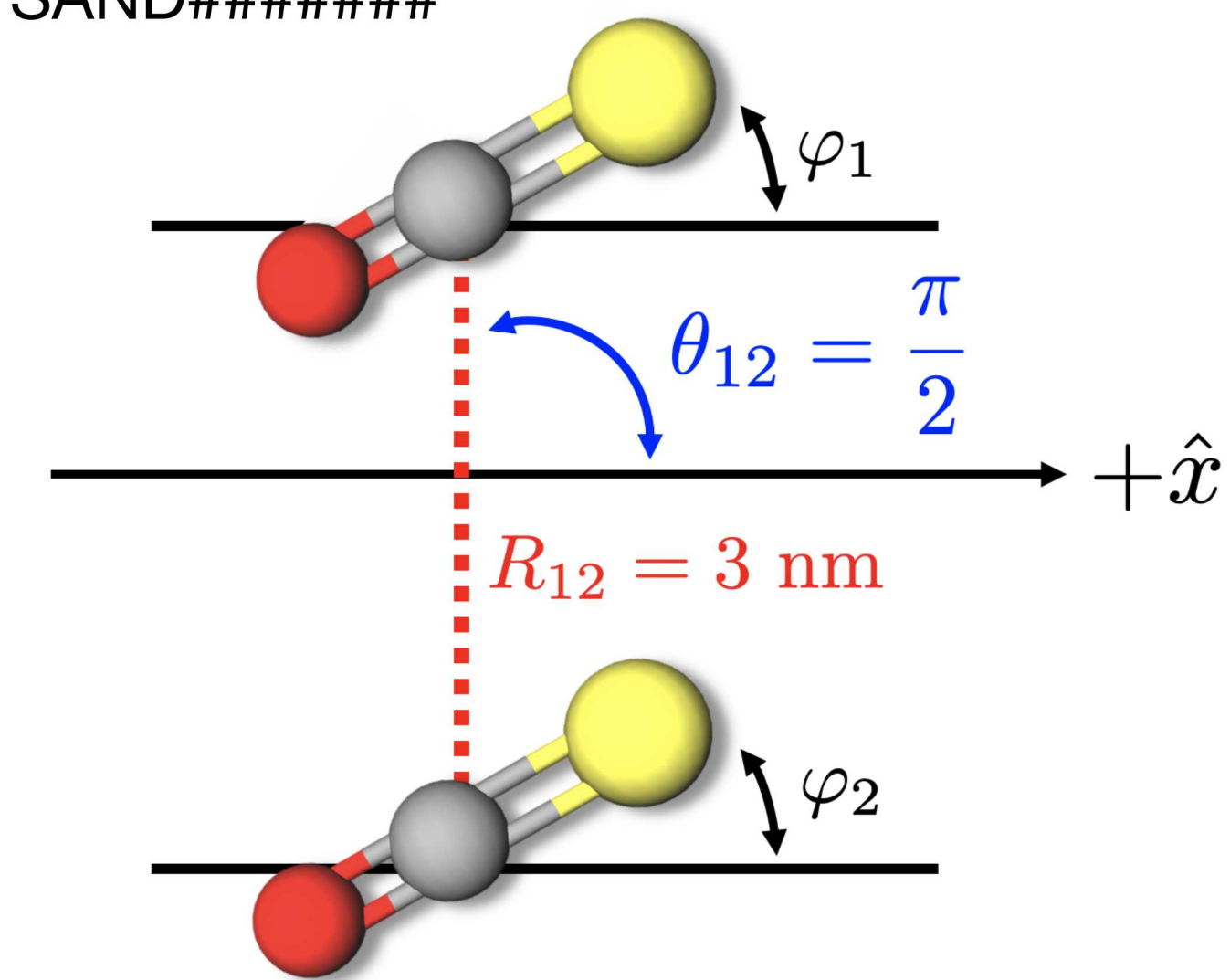
A. Guldberg and G. D. Billing, Chem. Phys. Lett **186**, 229 (1991).

Parameters $\{\theta_i\}$ taken to be amplitudes $\{a_i\}$, phases $\{\phi_i\}$, and detunings $\{\Delta_i\}$ of a mix of frequency components, and are updated using a genetic algorithm

$$f(t, p, a_1, \dots, a_k, \Delta_1, \dots, \Delta_k, \phi_1, \dots, \phi_k) = \sin^{1/p}\left(\frac{\pi t}{T}\right) \left(\sum_{i=1}^k a_i \cos((\omega_i + \Delta_i) - \phi_i) \right)$$

The system is represented in the harmonic oscillator eigenbasis and encoded using $N = 4$ qubits with initialization $|\psi(0)\rangle = |0\rangle|0\rangle|0\rangle|0\rangle$





Controlled orientation of 2 dipole-dipole coupled OCS rotors

Goal: control the rotational dynamics of two planar dipole-dipole coupled OCS rotors

$$H(t) = \sum_{j=1}^2 -\frac{\hbar^2}{2I} \frac{\partial^2}{\partial \varphi_j^2} + V_{12} - \mu(\cos \varphi_1 + \cos \varphi_2) f(t, \{\theta_i\})$$

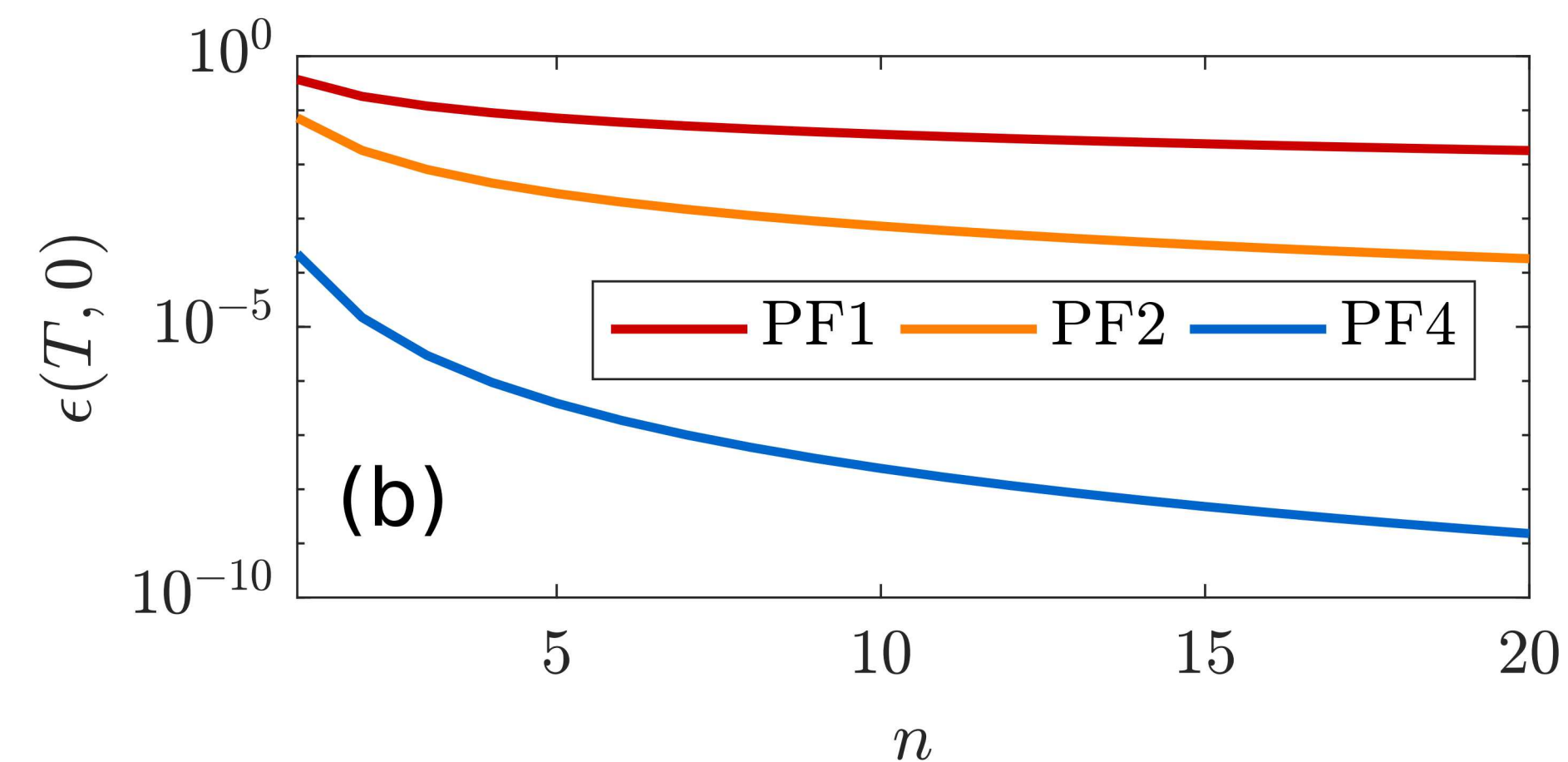
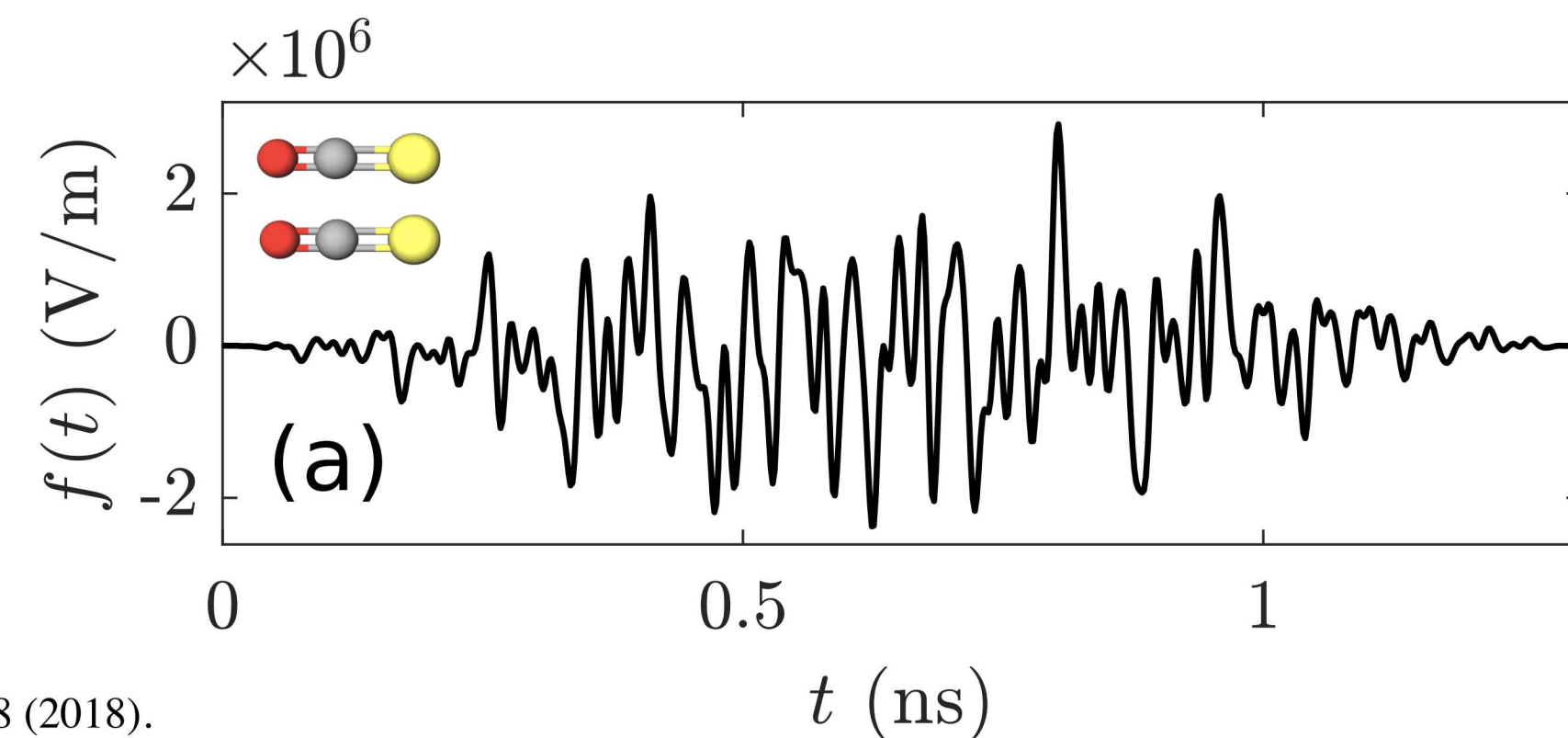
H. Yu, T.-S. Ho, and H. Rabitz, Phys. Chem. Chem. Phys. **20**, 13008 (2018).

such that they are oriented in the $+\hat{x}$ direction at time T

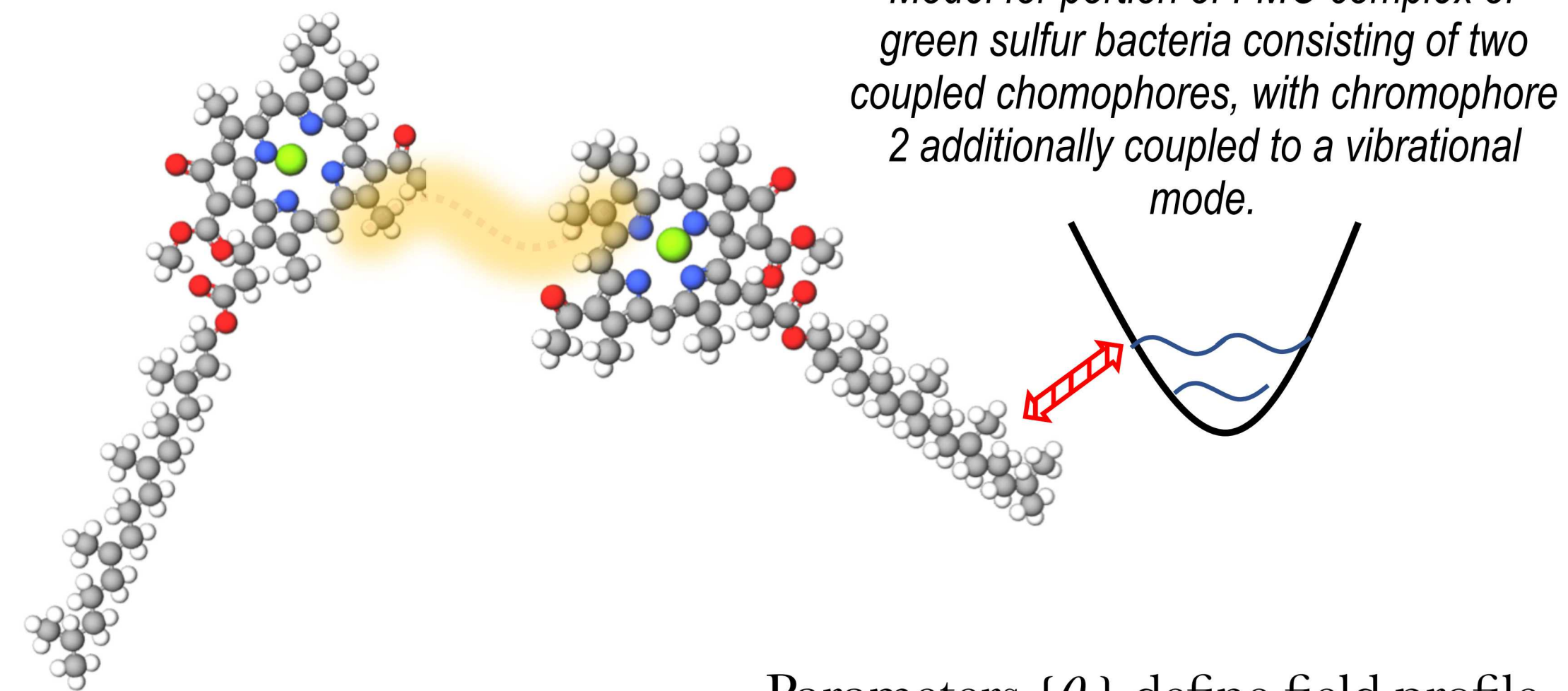
$$J[T, \{\theta_i\}] = -\langle \psi(T) | (\cos \varphi_1 + \cos \varphi_2) | \psi(T) \rangle$$

System modeled in free rotor eigenbasis and encoded using $N = 6$ qubits with initialization $|\psi(0)\rangle = |0\rangle |0\rangle \dots |0\rangle$

Parameters $\{\theta_i\}$ taken to be amplitudes, phases, and detuning of a mix of frequency components; Updated using a genetic algorithm



Controlled excitonic state preparation in light harvesting complex



Goal: prepare an excitation on the second chromophore when the vibrational mode is initially in a thermal state $\sum_{v=0}^{v_{max}} c_v |v\rangle\langle v|$ at 300K, where $c_v = \frac{1}{\text{tr}\{e^{-\beta H_v}\}} e^{-\beta v}$.

$N = 5$ qubits are used for the encoding, and the control objective functional is given by

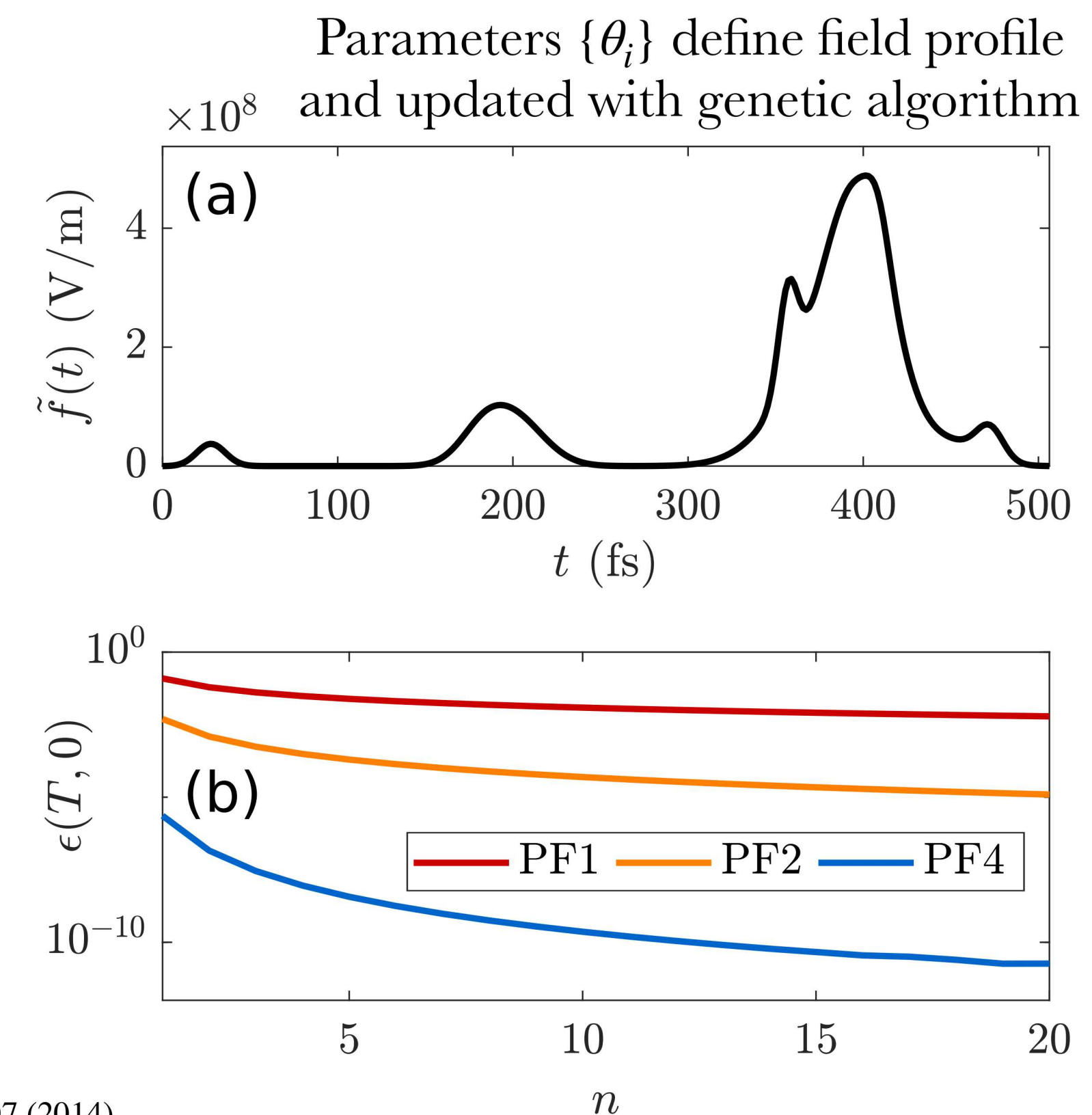
$$J[T, \{\theta_i\}] = \sum_{v=0}^{v_{max}} c_v J_v[T, \{\theta_i\}]$$

where $J_v[T, \{\theta_i\}] = 1 - \langle \psi(T) | (|g_1 e_2\rangle\langle g_1 e_2|) I_v | \psi(T) \rangle$, $|\psi(0)\rangle = |0\rangle |0\rangle |v\rangle$

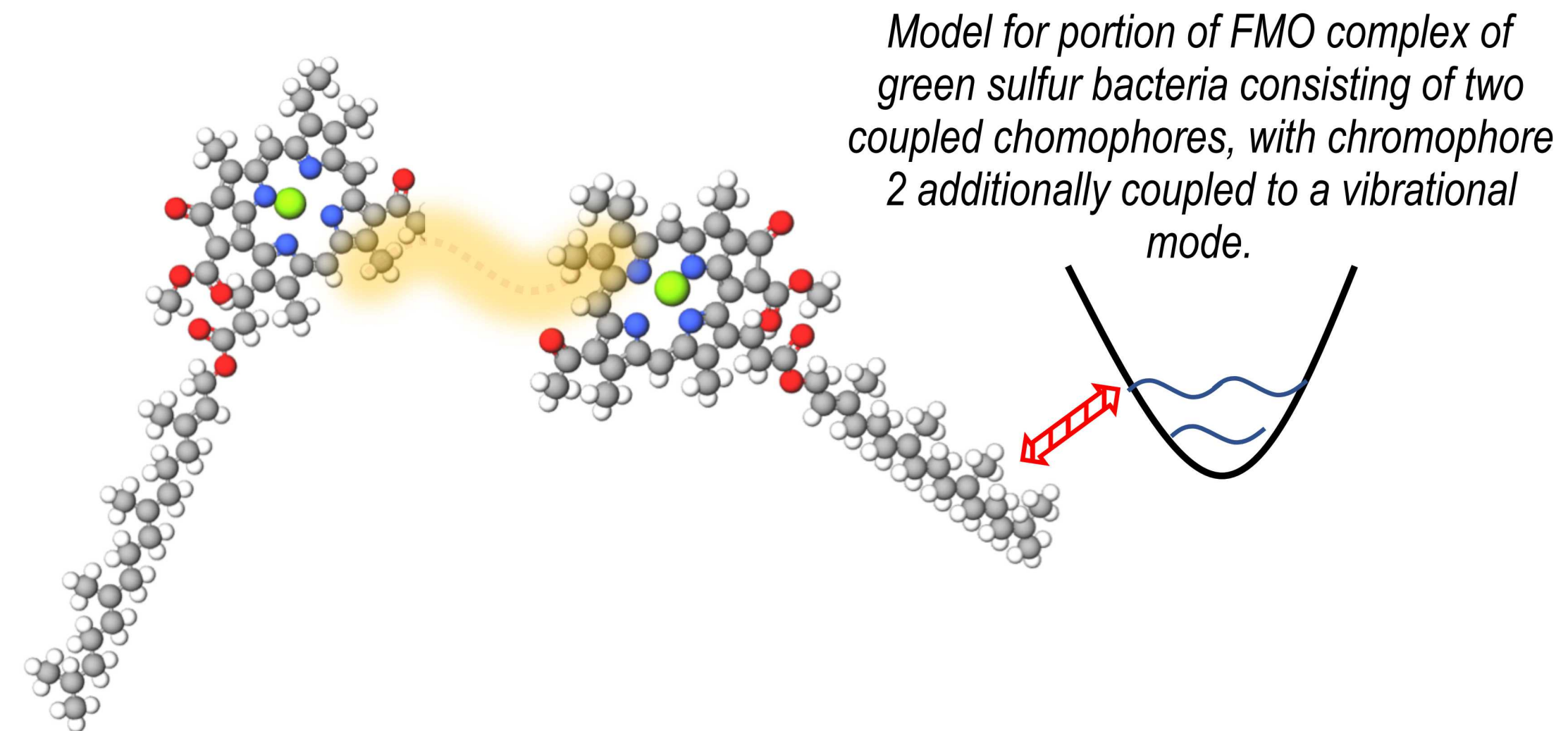
and the Hamiltonian is given by,

$$H(t) = \sum_{j=1}^2 E_j b_j^\dagger b_j + J(b_1^\dagger b_2 + b_2^\dagger b_1) + \nu a^\dagger a + g(b_2^\dagger b_2)(a + a^\dagger) + \sum_{j=1}^2 (\mu_j(b_j + b_j^\dagger))f(t, \{\theta_i\})$$

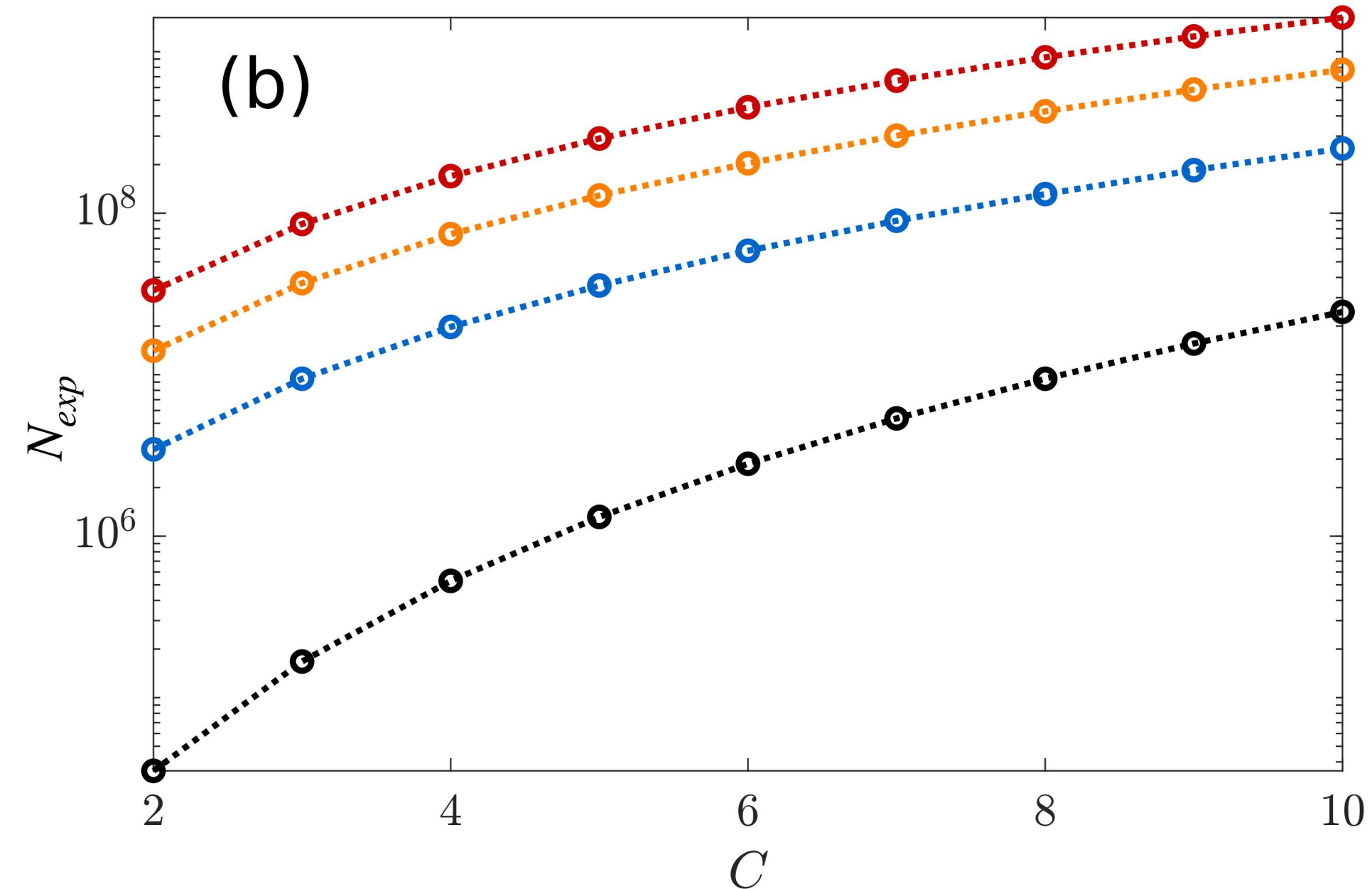
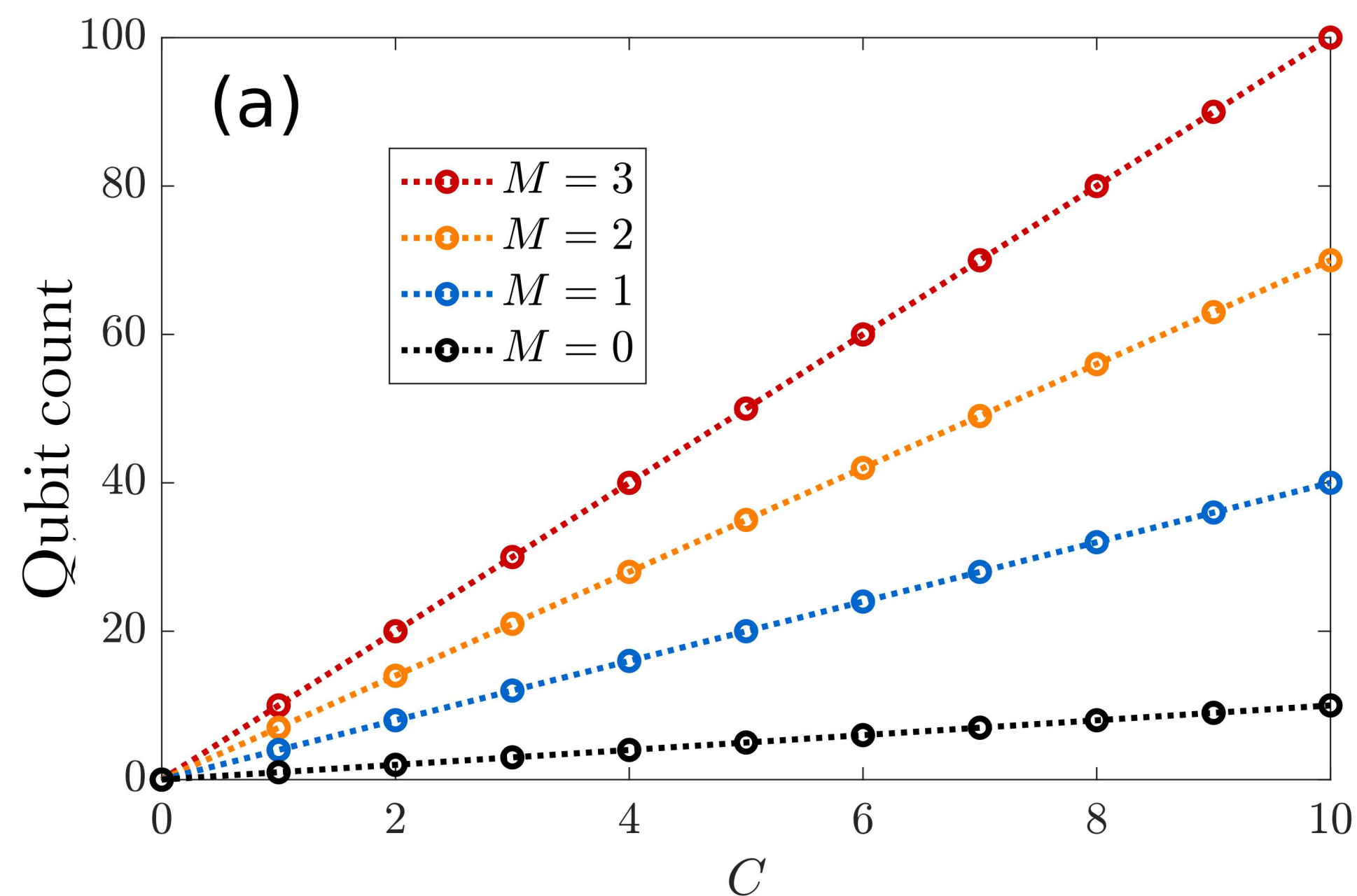
Model: S. Hoyer, et al, New J. Phys **16**, 045007 (2014).



Controlled excitonic state preparation in light harvesting complex

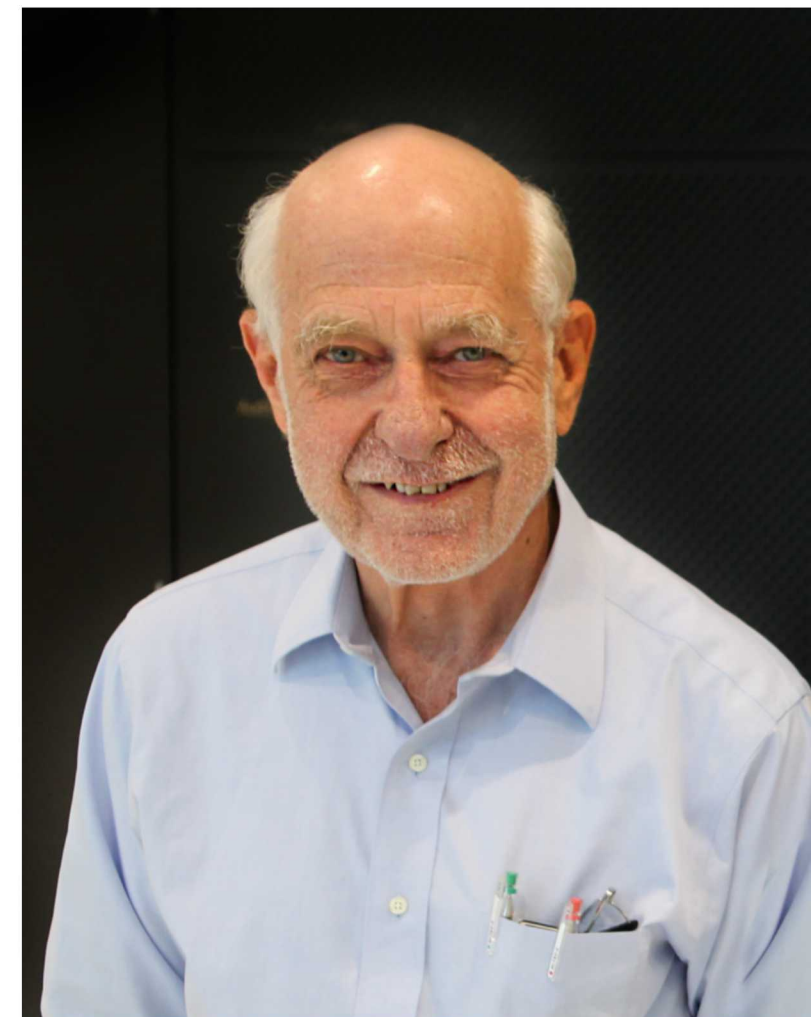


Resource analysis for simulating the dynamics of C coupled chromophores, each coupled to M vibrational modes, on a quantum computer

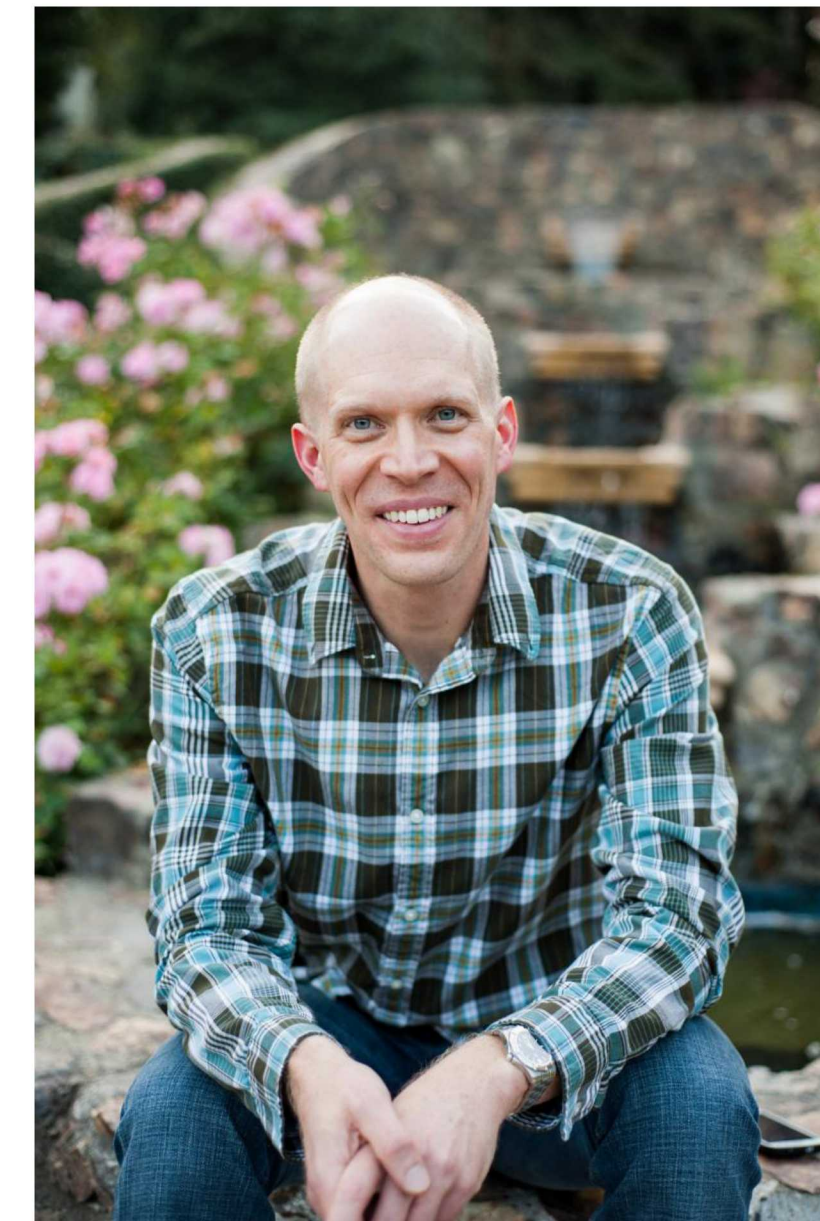


D. W. Berry, G. Ahokas, R. Cleve, and B. C. Sanders, Commun. Math. Phys. **270**, 359 (2007).

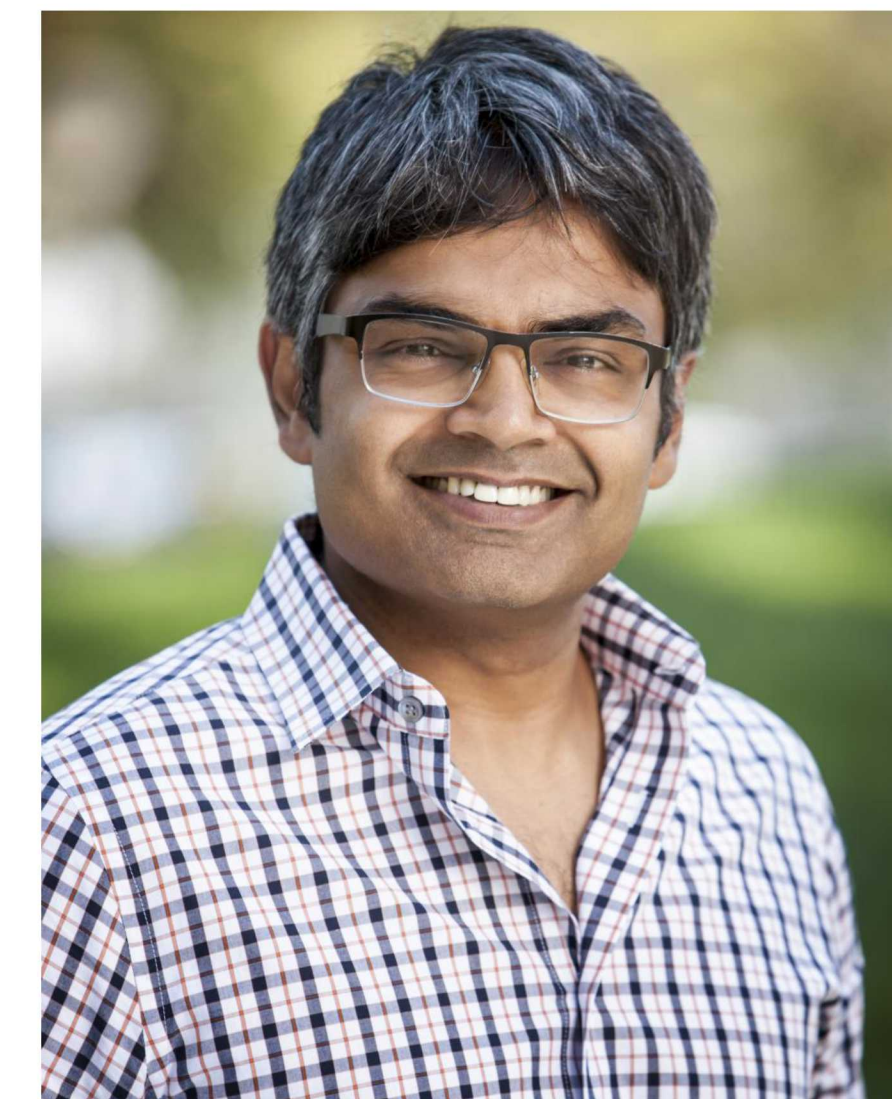
Thank you



**Herschel
Rabitz**



**Matthew
Grace**



**Mohan
Sarovar**

