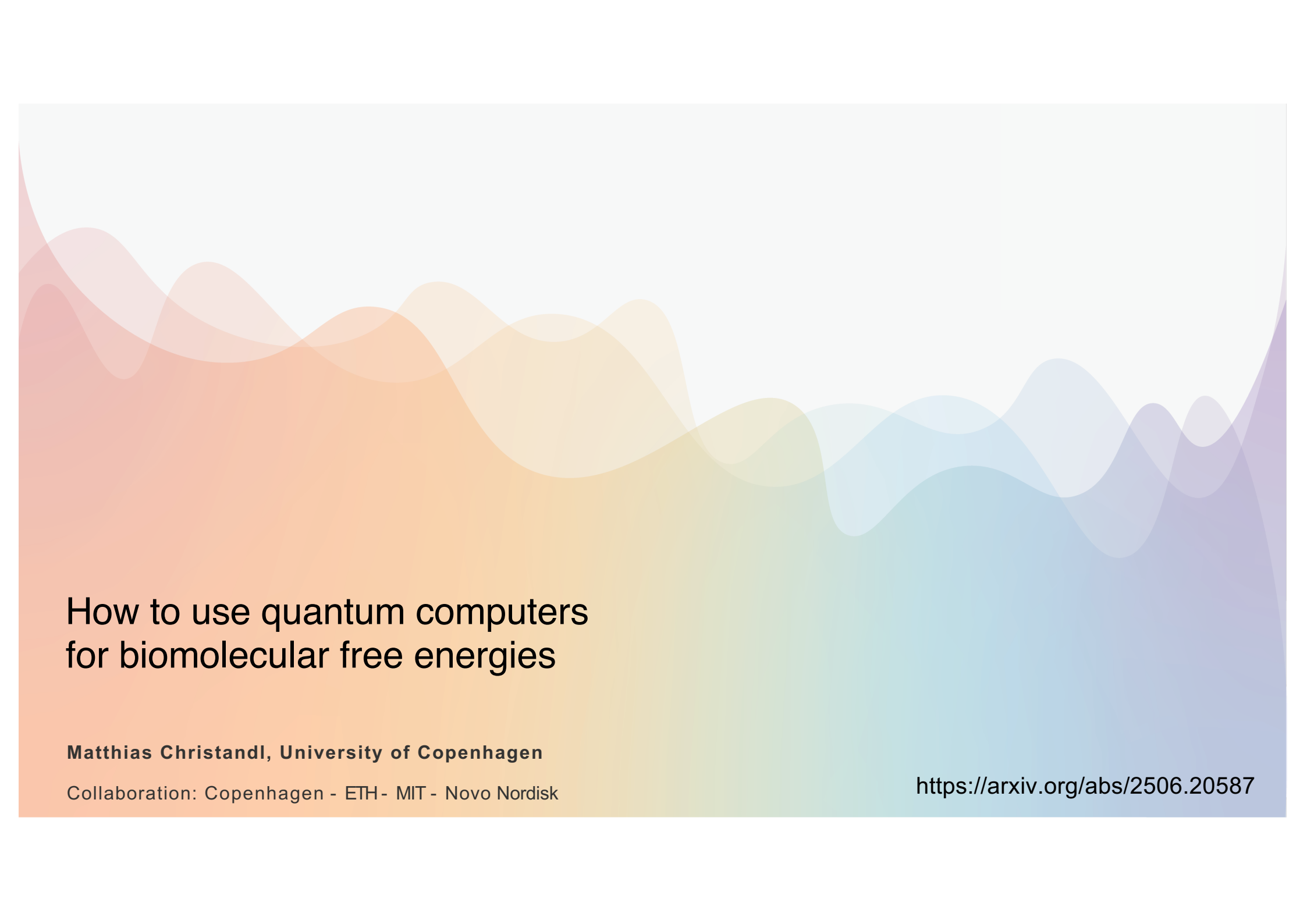




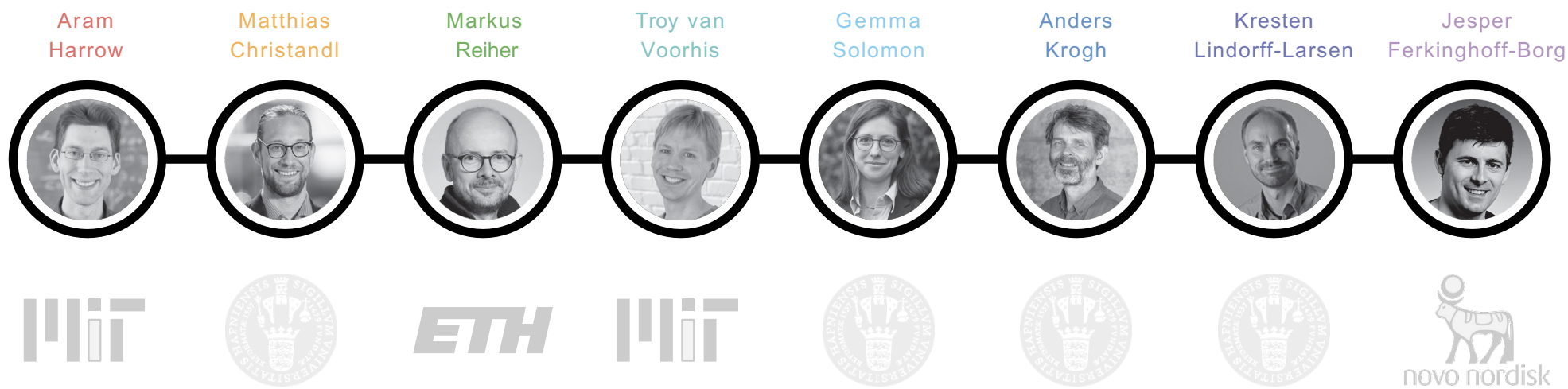
How to use quantum computers for biomolecular free energies

Matthias Christandl, University of Copenhagen

Collaboration: Copenhagen - ETH - MIT - Novo Nordisk

<https://arxiv.org/abs/2506.20587>

The Group Leaders



Quantum Algorithms

Quantum Chemical Simulations

Classical Chemical Simulations

Protein Modeling

Human Health Science

NQCP
NIELS BOHR INSTITUTE


Quantum for Life



novo nordisk
foundation
Benefitting people and society

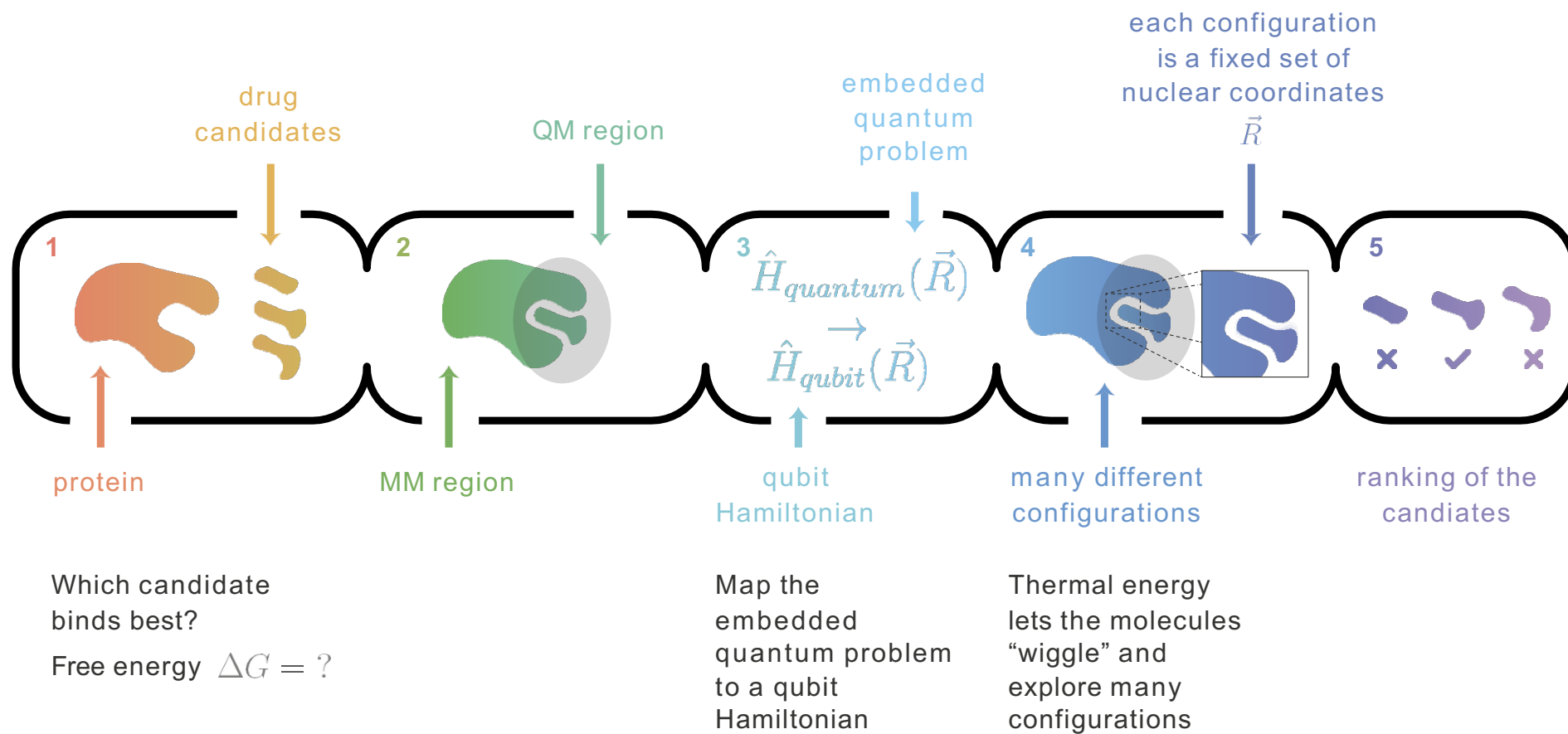
 **Quantum for Bio**

leap^w

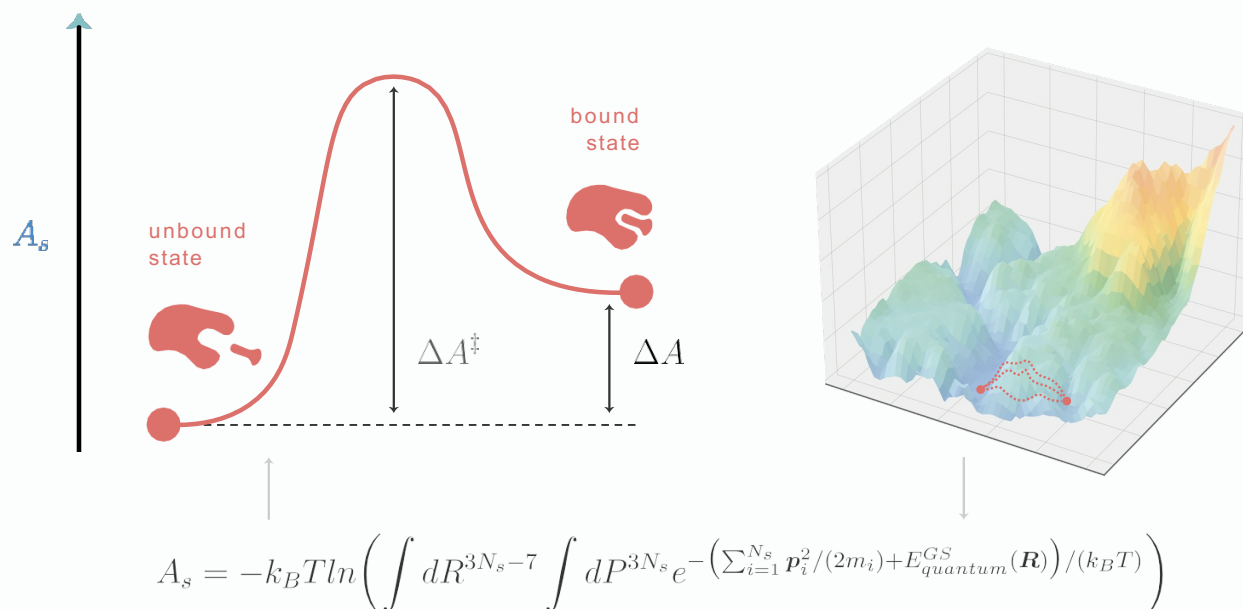
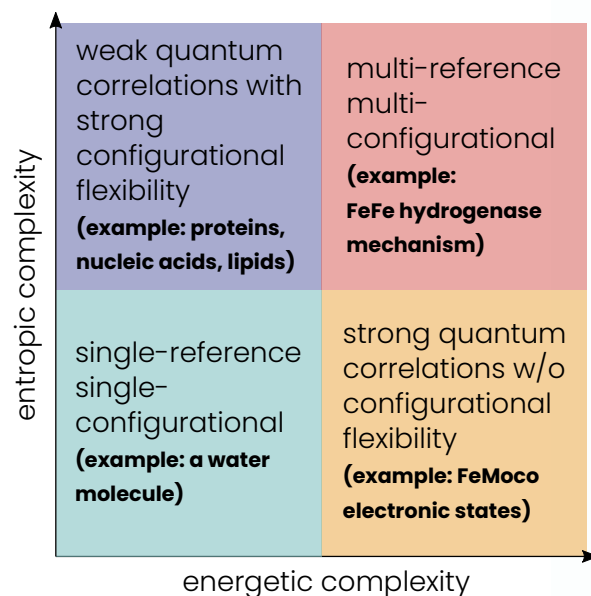
The Team



The Project



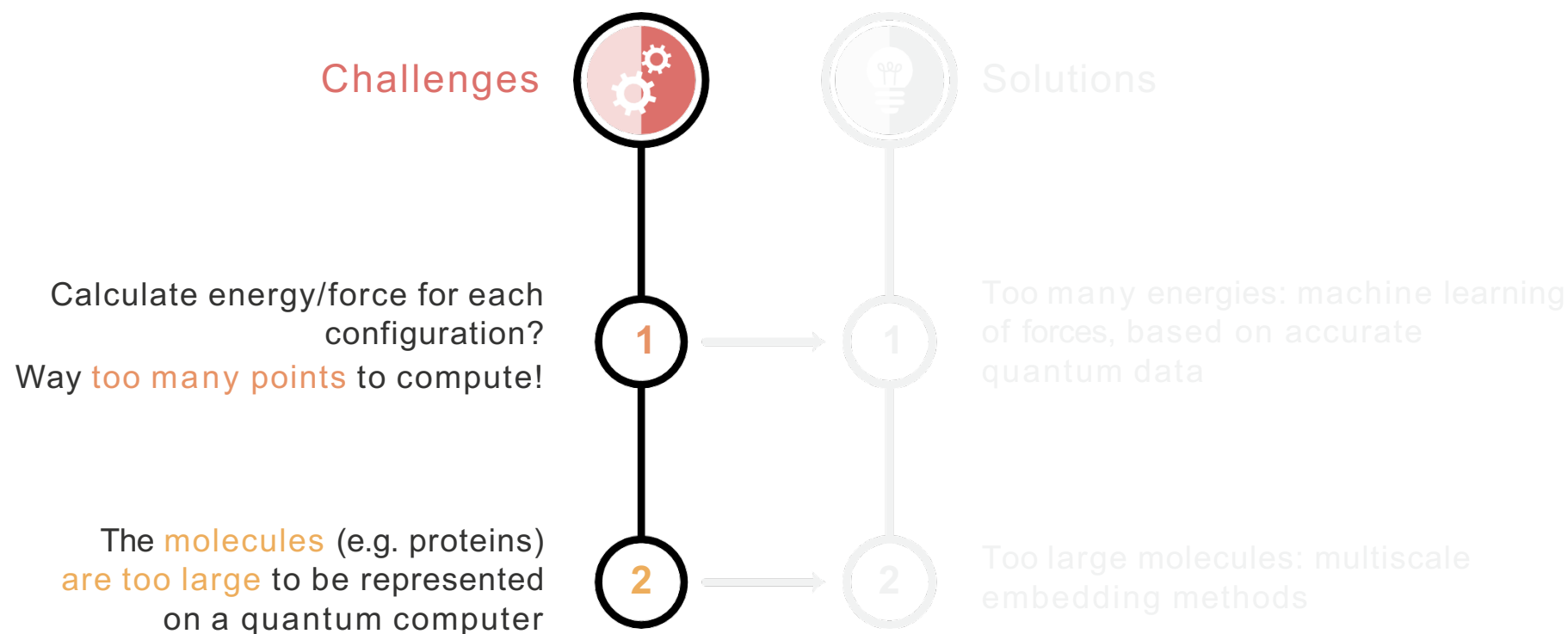
Energy versus entropy



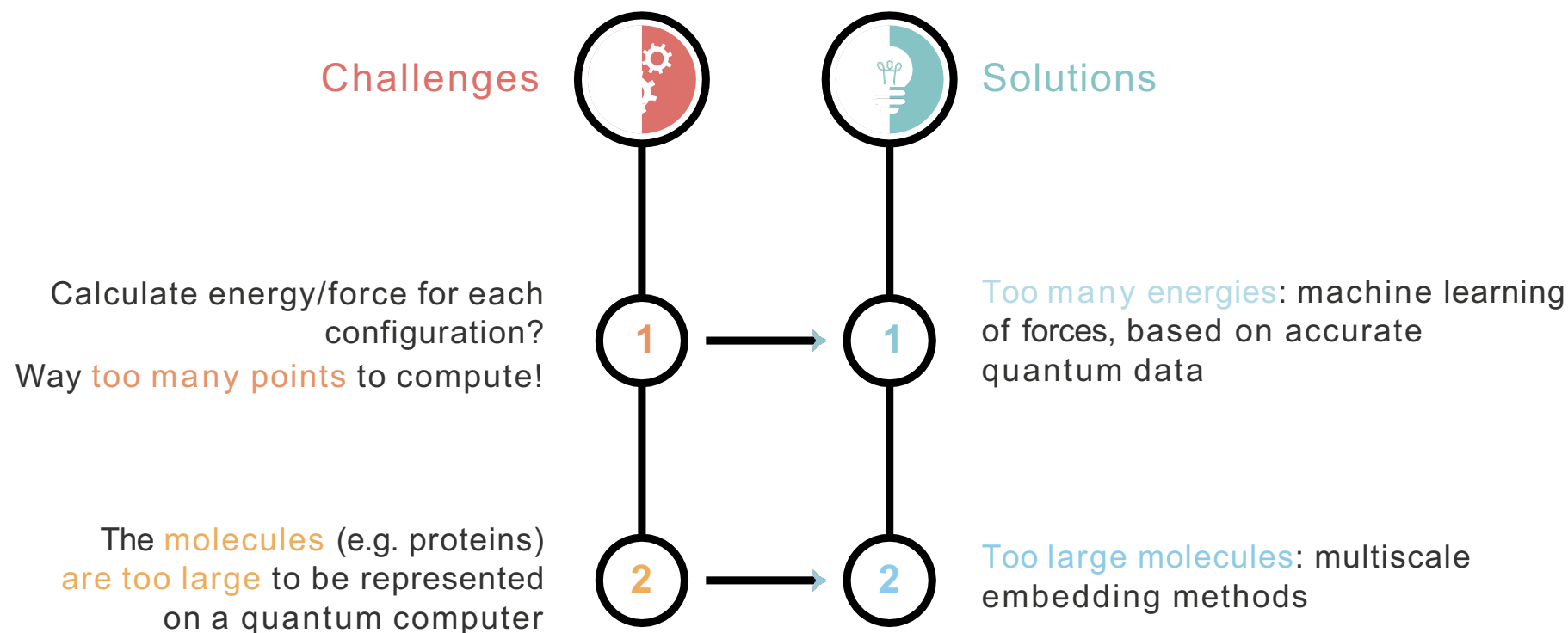
Biomolecular simulation quadrangle:
Entropic versus energetic complexity

Free energy: sample from the potential energy surface
Current approach: solve Newton's equations using heuristic forces
Quantum promise: compute the true forces/energies for electronic interaction

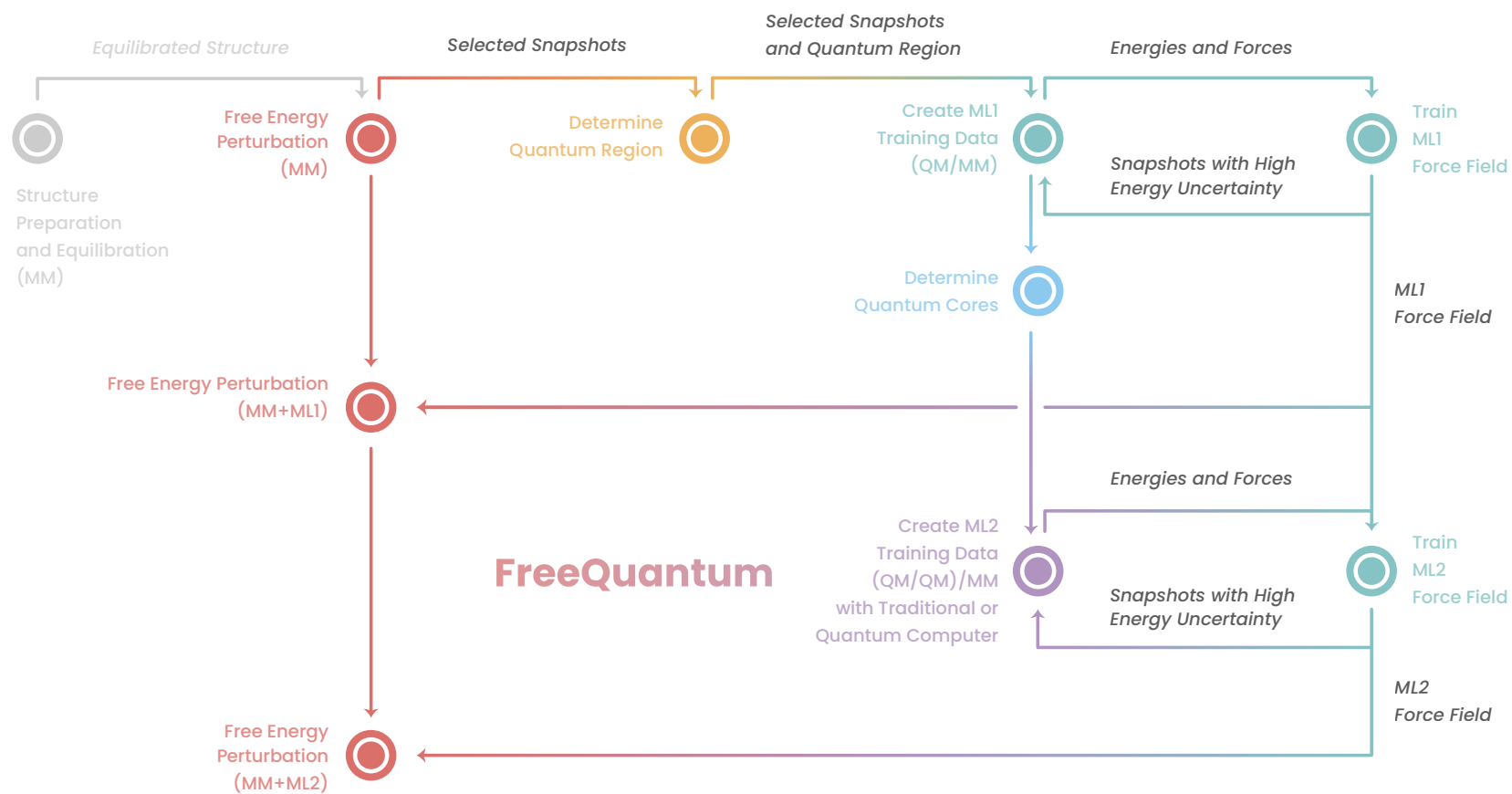
Quantum Computing for Free Energy Sampling?



Quantum Computing for Free Energy Sampling?



Our Computational Pipeline



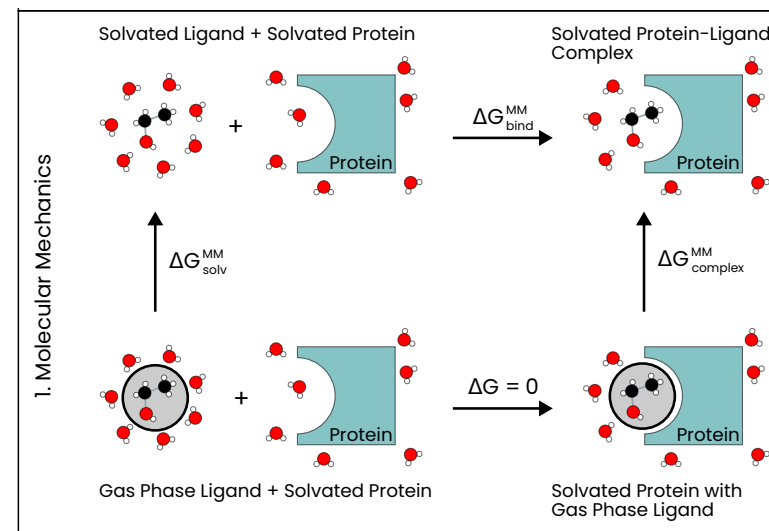
Free Energy Perturbation

- Partition function $Z = \int d^3N_{\text{nuc}} R \text{tr} \exp(-\beta H(R))$

- Free energy difference $\Delta G_{\text{binding}} = G_{\text{bound}} - G_{\text{unbound}}$

- Telescoping sum along coordinate to make sampling problems more similar (e.g. reaction coordinate or alchemical)

$$= \sum_{k=1}^{s-1} \underbrace{G_{\lambda_k} - G_{\lambda_{k+1}}}_{\Delta G_{\lambda_k}}$$



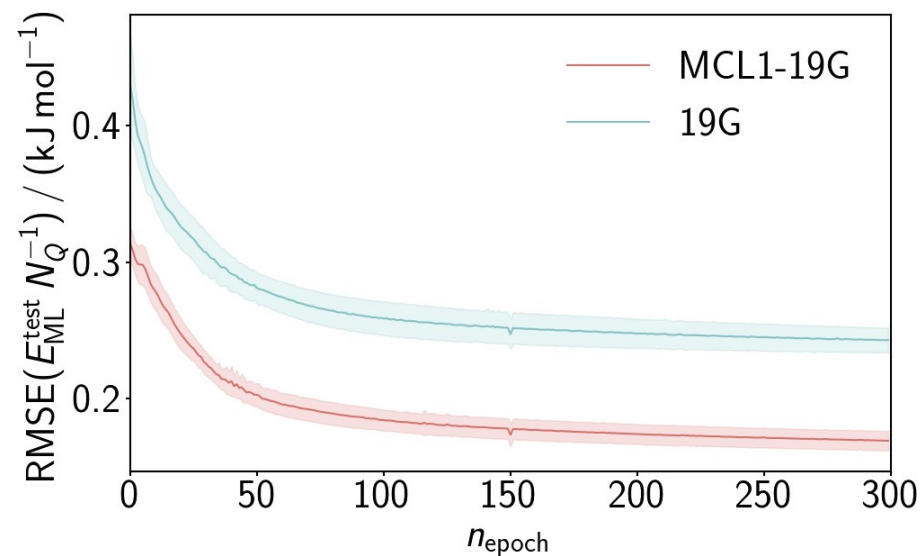
- Alchemical: switch interaction off/on $H_{\lambda} = H_{\text{protein}} + H_{\text{ligand}} + \lambda H_{\text{interaction}}$

$$\Delta G_{\text{solvated binding}} = \underbrace{G_{\text{solvated complex}} - (G_{\text{solvated protein}} + G_{\text{ligand}})}_{\Delta G_{\text{partially solvated binding}}} + \underbrace{(G_{\text{ligand}} - G_{\text{solvated ligand}})}_{-\Delta G_{\text{ligand solvation}}}$$

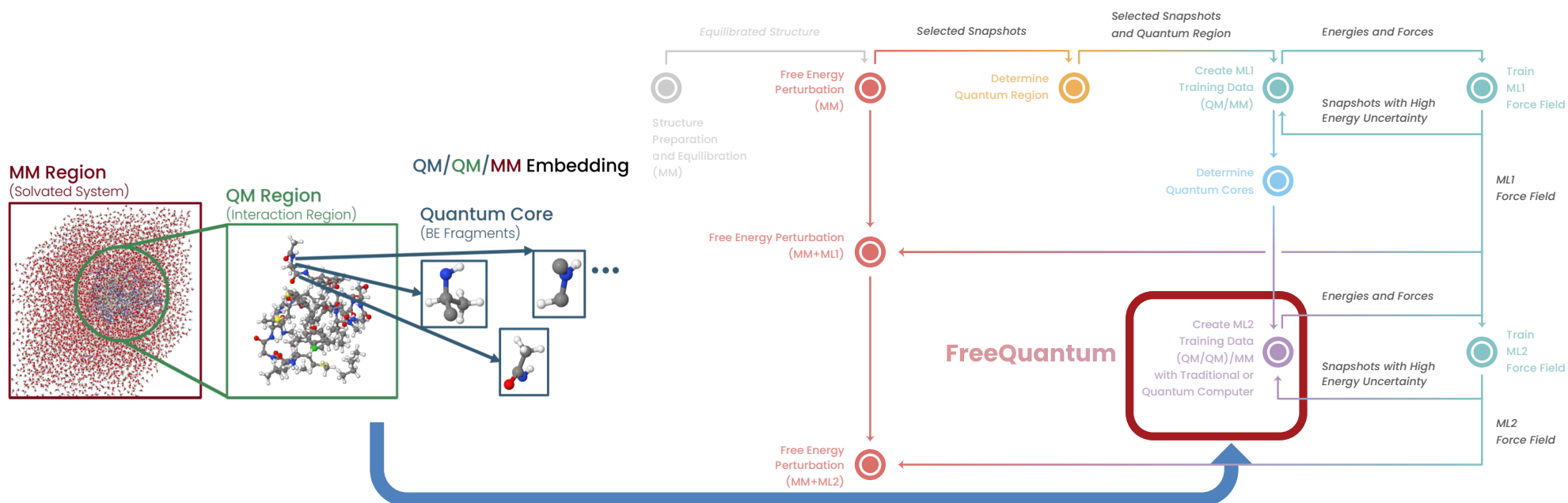
- Then telescoping with reaction coordinate for each part

Machine Learning

- Machine-learning potentials are typically trained on energies and forces
- Energies alone are sufficient using transfer learning
- Transfer learning allows to leverage best the few energy data points available from quantum computing hardware



Quantum Embedding



$$H_{\text{quantum region}} = H_{\text{electronic, quantum region}} + V_{\text{interaction, MM}}$$

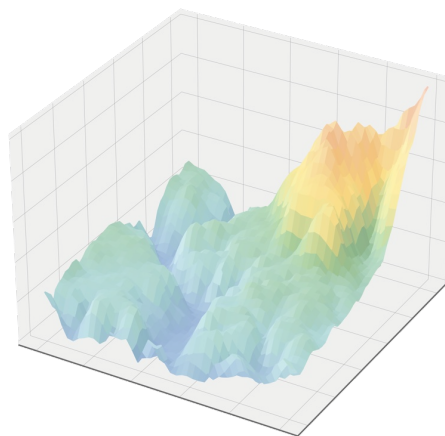
$$H_{\text{quantum core}} = H_{\text{electronic, quantum core}} + V_{\text{interaction, DFT}}$$

Traditional Quantum Engine

Full problem $\hat{H}(t)\Psi(\mathbf{r}, \mathbf{R}, t) = i\hbar\partial_t\Psi(\mathbf{r}, \mathbf{R}, t)$

Stationary states $\hat{H}\Psi_n(\mathbf{r}, \mathbf{R}) = E_n\Psi_n(\mathbf{r}, \mathbf{R})$

Born-Oppenheimer $\Psi_{\text{nuc}}(\mathbf{R}) \times \Psi_{\text{ele}}^{\mathbf{R}}(\mathbf{r})$



Exponential explosion

Hierarchy of approximations

DMRG/Tensor Network as proxy for Full-CI

Computationally challenging

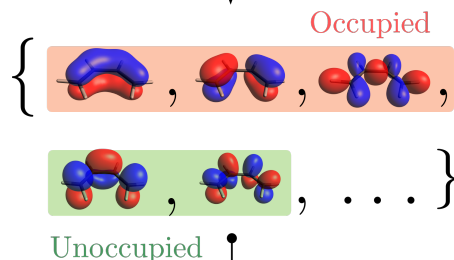
No accuracy guarantees

Many-body wave function

$$\Psi(\mathbf{r}_1, \dots, \mathbf{r}_{N_{\text{ele}}})$$

Superposition of configurations

$$|\Psi\rangle = C_0|\Psi_0\rangle + \sum_i \sum_a C_i^a |\Psi_i^a\rangle + \sum_{ij} \sum_{ab} C_{ij}^{ab} |\Psi_{i,j}^{a,b}\rangle + \dots \quad \text{Full CI}$$



One-particle functions

$$\{\varphi_1(\mathbf{r}), \varphi_2(\mathbf{r}), \dots, \varphi_{N_{\text{orb}}}(\mathbf{r})\}$$

$$|\varphi_1 \dots \varphi_i \dots \varphi_a \dots \varphi_b \dots \varphi_{N_{\text{ele}}}\rangle = |\Psi_{i,j}^{a,b}\rangle \quad \text{CISD}$$

Double substitution

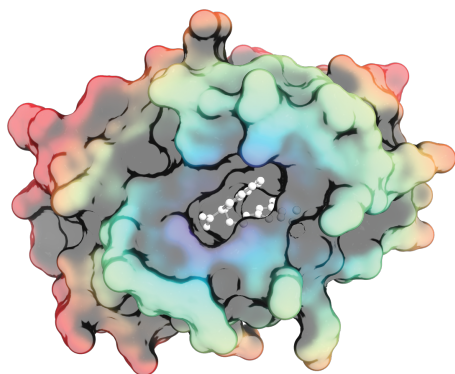
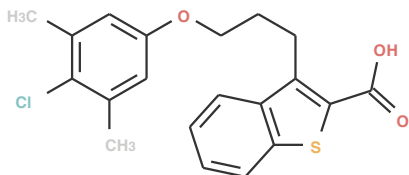
$$|\varphi_1 \dots \varphi_a \dots \varphi_j \dots \varphi_k \dots \varphi_{N_{\text{ele}}}\rangle = |\Psi_i^a\rangle \quad \text{CIS}$$

Single substitution

$$|\varphi_1 \dots \varphi_i \dots \varphi_j \dots \varphi_k \dots \varphi_{N_{\text{ele}}}\rangle = |\Psi_0\rangle \quad \text{HF}$$

Reference configuration

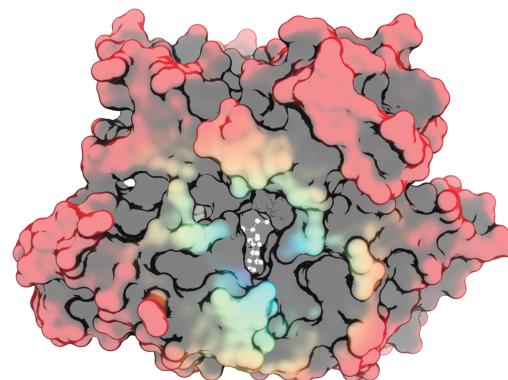
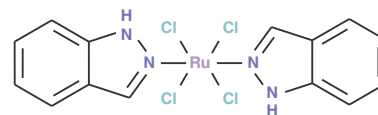
The Biological Systems



The MCL1-19G complex
(the “model complex”)

Benchmark data available

Used for testing purposes



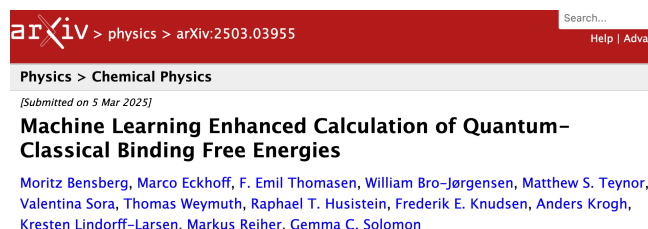
The GRP78-NKP1339 complex
(the “Ruthenium complex”)

No benchmark data available

Electronic structure challenge

Full Computational Run

- The FreeQuantum pipeline is based on first principles
- Benchmarking: The MCL1-19G model complex (protein MCL1 and small molecule 19G) demonstrates that our pipeline yields reliable results

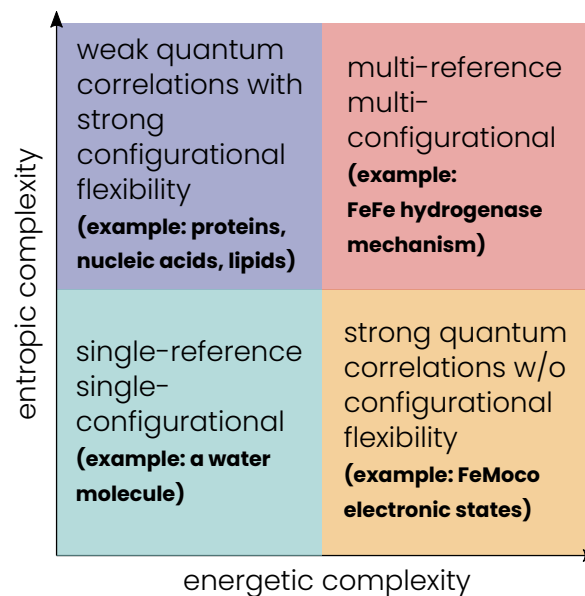


- It can be applied in cases where standard MM models are not available
- For this, the GRP78-NKP1339 complex (protein GRP78 and small molecule NKP1339) serves as an example

Method	MCL-1/19G
MM	-37.5 ± 0.4 [20]
PBE/MM	-35.3 ± 1.8 [20]
UMP2/PBE/MM	—
LCCSD(T)/PBE/MM	-37.2 ± 1.0 [21]
UCCSD(T)/PBE/MM	—
NEVPT2/PBE/MM	—
Experiment	-37.3 ± 0.1 [43]

The potential of quantum computing

- Energy computation of quantum cores carried out by quantum computers
- Advantages: accuracy guarantees, more precise for larger and more correlated systems
- With the FreeQuantum pipeline we built a computational framework that allows to slot-in the quantum computer
- Ready for quantum advantage, when quantum hardware is



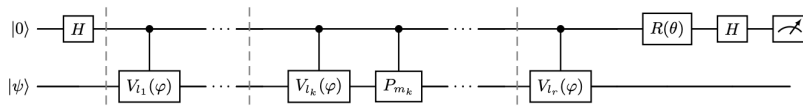
Biomolecular simulation quadrangle:

Entropic versus energetic complexity

Quantum Phase Estimation



- The main cost of phase estimation derives from the cost of Hamiltonian simulation
- Chemistry: an unstructured Hamiltonian with many terms which leads to deep circuits
- We use single-ancilla quantum phase estimation (QPE)



- Partially randomized product formulas (Trotter) need no ancillas and reduce circuit depth

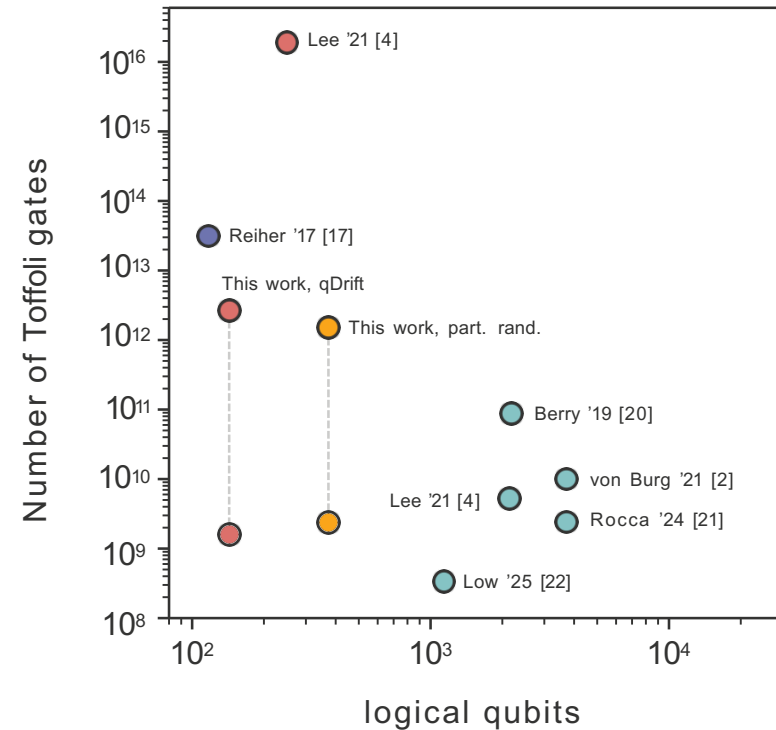


Figure: FeMoCo benchmark

$$|\Psi(t)\rangle \approx \prod_{i=1}^N \left(e^{-i\hat{H}_1 t/N\hbar} e^{-i\hat{H}_2 t/N\hbar} \right) |\Psi(0)\rangle$$

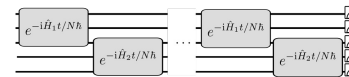


Figure: Trotter reminder

Quantum Resources

- Partially randomized product formulas need no ancillas and reduce circuit depth
- Runtime (Trotter constant) estimate using novel software on large GPU cluster

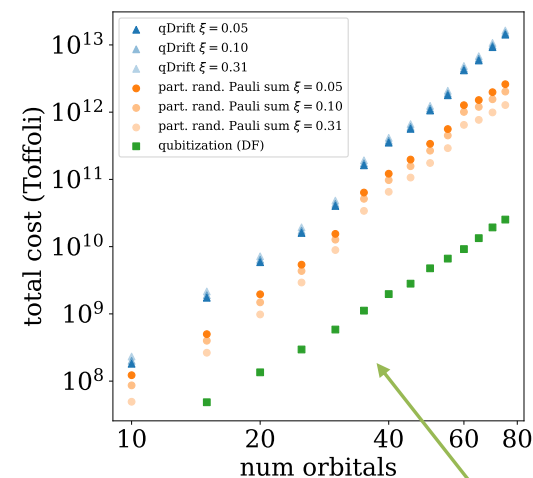
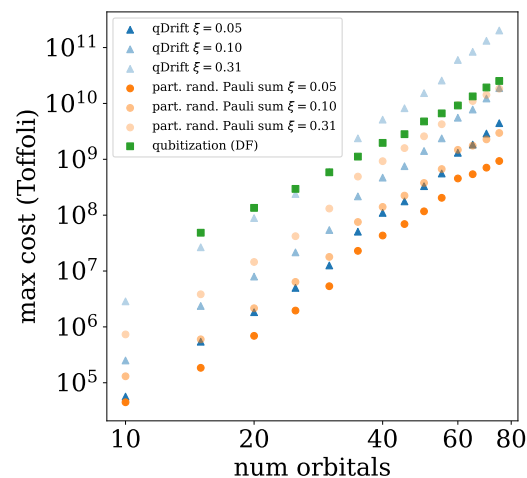


- It is possible to prepare initial states with high ground state overlap

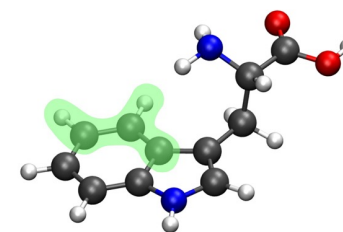
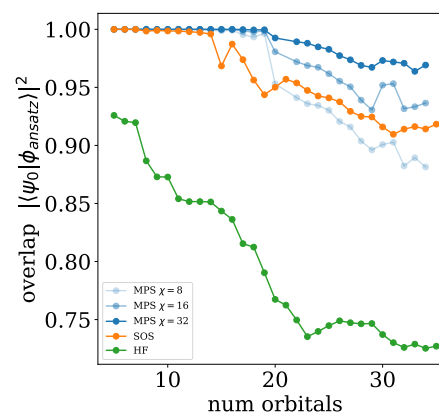
PRX LIFE 3, 013003 (2025)

High Ground State Overlap via Quantum Embedding Methods

Mihael Erakovic^{1,*}, Freek Witteveen^{2,†}, Dylan Harley², Jakob Günther², Moritz Bensberg¹, Oinam Romesh Meitei³, Minsik Cho³, Troy Van Voorhis³, Markus Reiher^{1,‡} and Matthias Christandl^{2,§}



substantial
#ancilla qubits



Conclusion

Developed a complete, modular, and autonomous computational pipeline for free energy calculations.

Compatible with traditional quantum chemistry and future fault-tolerant quantum computers.

Adapts to computational resources; modules can be exchanged or upgraded.

Ruthenium drug–protein complex (open-shell spin doublet) demonstrated feasibility with HPC.

Impact: Multilayer embedding + ML reduces quantum region size → makes quantum computing for free energy feasible.

Opens path toward quantum advantage in biology; FreeQuantum is open source and free.

How to use quantum computers for biomolecular free energies

<https://arxiv.org/abs/2506.20587>

