



An optimal quantum sampling regression algorithm for variational eigensolving in the low qubit number regime

2021 Chicago Quantum Exchange Workshop

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| arXiv:2012.02338

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Variational Quantum Eigensolver (VQE)

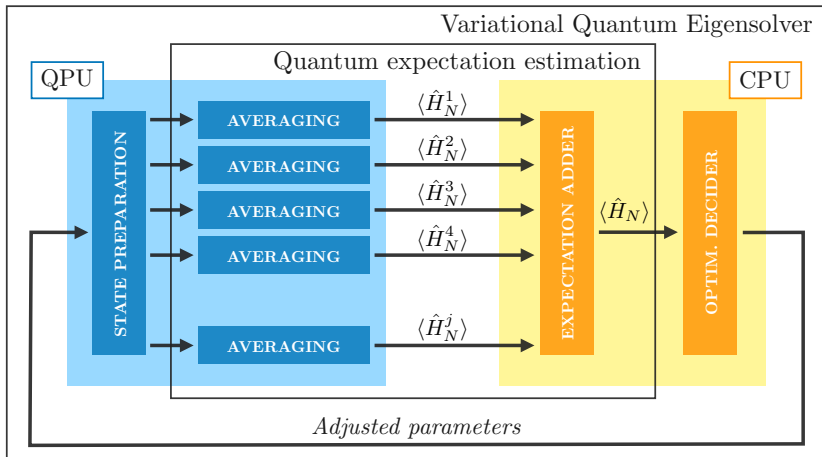
VQE algorithm

QSR algorithm

Complexity

Applications

Banchmarking



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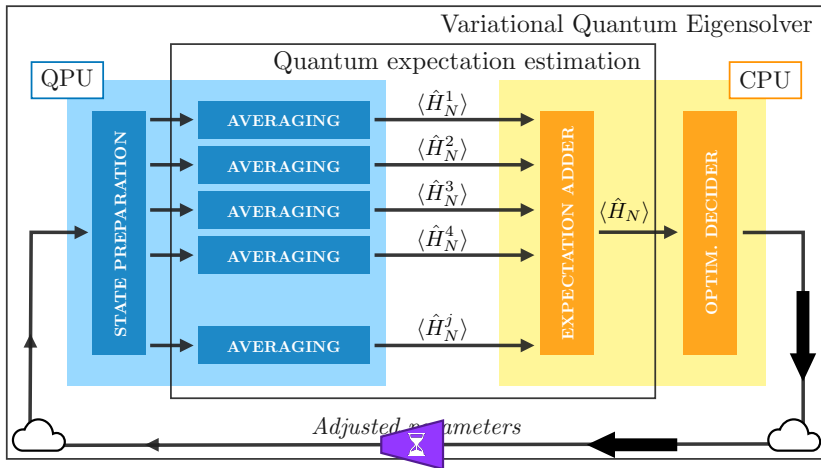
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Quantum Sampling Regression (QSR)

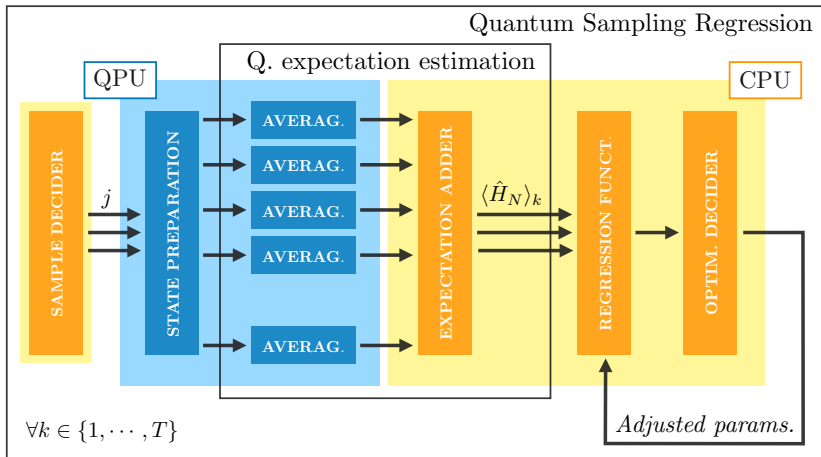
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Quantum Sampling Regression (QSR)

VQE algorithm

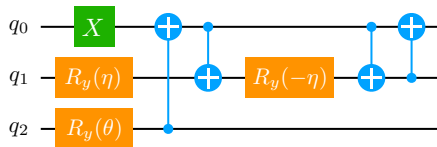
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- From the **topology of the quantum circuit** in charge of state preparation, we can infer a frequency bound.
- **Fourier analysis** then allows to fully reconstruct the expectation value function.
- Through the **Nyquist-Shannon sampling theorem** we can show that our sampling technique is optimal.



Theorem (Nyquist-Shannon)

If a function $h(\theta)$ contains no angular frequencies higher than ω_S , it is completely determined by giving its ordinates at a series of points $1/2\omega_S$ apart: $\omega_{\text{sampling}} > 2\omega_S$.



Low qubit number regime

VQE algorithm

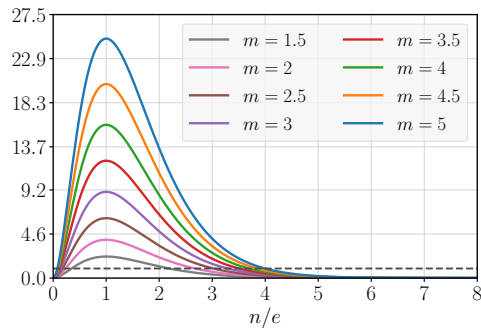
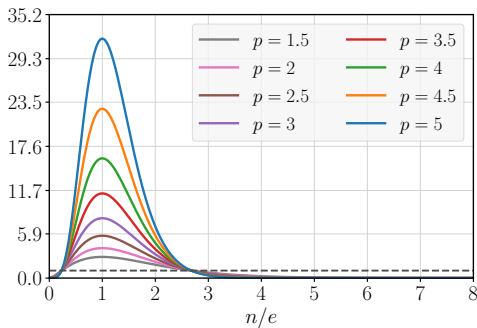
QSR algorithm

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■ Algorithmic complexity model: $\frac{\text{VQE}}{\text{QSR}} = \left(mn2^{-n/r}\right)^p$



Low qubit number regime

VQE algorithm

QSR algorithm

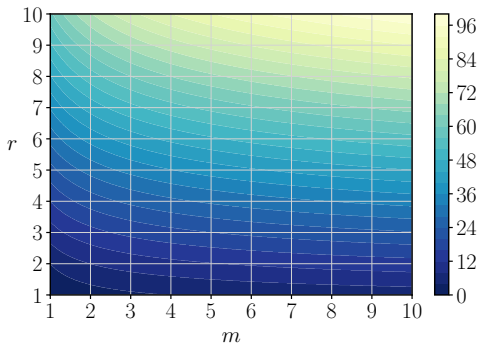
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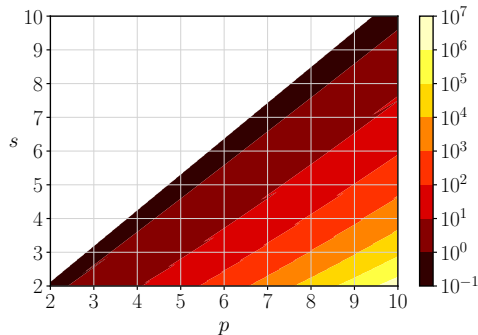


Threshold



$$a \triangleq \left\lceil -\frac{r}{\ln 2} W_{-1} \left(-\frac{\ln 2}{mr} \right) \right\rceil$$

Efficiency



$$E \approx \frac{1}{as \ln 2} \left(\frac{m}{s \ln 2} \right)^p \Gamma(p+1, s \ln 2, as \ln 2)$$

Applications

VQE algorithm

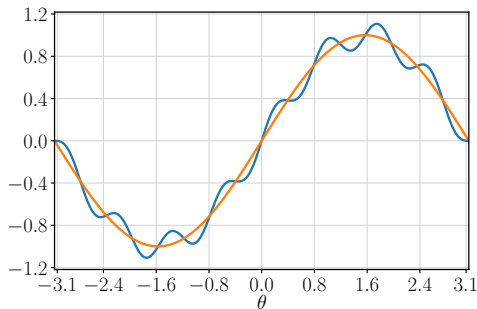
QSR algorithm

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- **Oversampling** to attain higher precision.
- **Undersampling** to boost performance and get rid of small-wavelength oscillations leading to burdensome local minima.
- VQE low-resolution start-up **supplement**.
- **Proxy** to transition between simulators and real devices.
- Improve convergence by removing the stochastic nature of the quantum expectation value function.
- Avoid the exponential matrix formulation in classical computation.



Banchmarking (arXiv:1801.03897)

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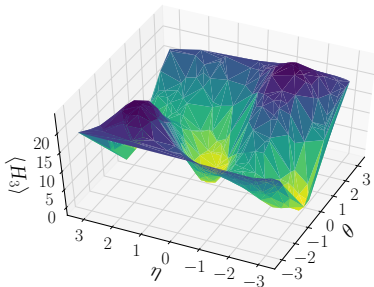
VQE

DEUTERON BINDING ENERGY:

Minimum 'E3 = -2.0513 (MeV)',

'[theta, eta] = [0.2819, 0.3040] (rad)',

» ERROR = 0.3%



183 samples \rightarrow 183 queries

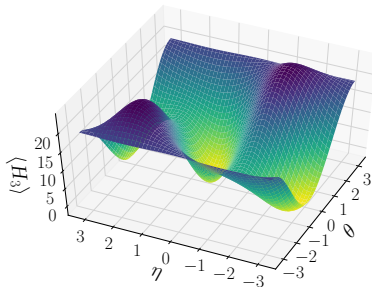
QSR

DEUTERON BINDING ENERGY:

Minimum 'E3 = -2.0509 (MeV)',

'[theta, eta] = [0.2688, 0.3631] (rad)',

» ERROR = 0.2%



25 samples \rightarrow 1 query



Thanks

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