

A Quantum Adaptive Importance Sampling Algorithm for Multidimensional Integration

Paper

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1. Introduction and Objectives

Precision in Perturbative QFT:

Numerical integration of computationally demanding multi-loop Feynman diagrams and phase-space integrals.

Classical Approach: Adaptive Importance Sampling (VEGAS):

Effective for sharp peaks. Allocates samples via adapting a grid. Mitigates the curse of dimensionality with a separable probability density function (PDF).

Quantum Adaptive Importance Sampling (QAIS):

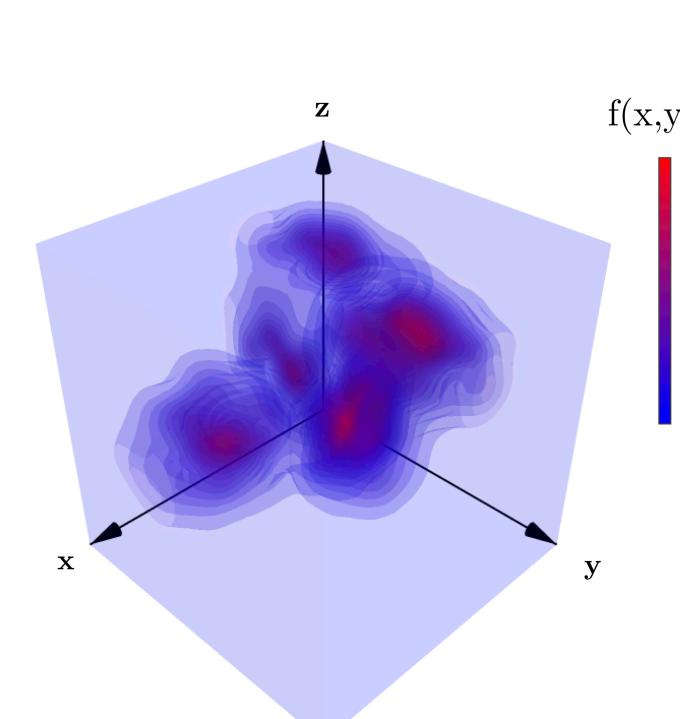
Exploit the exponentially large Hilbert space of a parameterised quantum circuit (PQC) to load and sample from a non-separable proposal PDF.

Why Importance Sampling (IS) with Quantum Computing?

IS targets small, high-impact regions. Direct sampling with a quantum circuit is efficient when the answer lies in a small corner of the full Hilbert space.

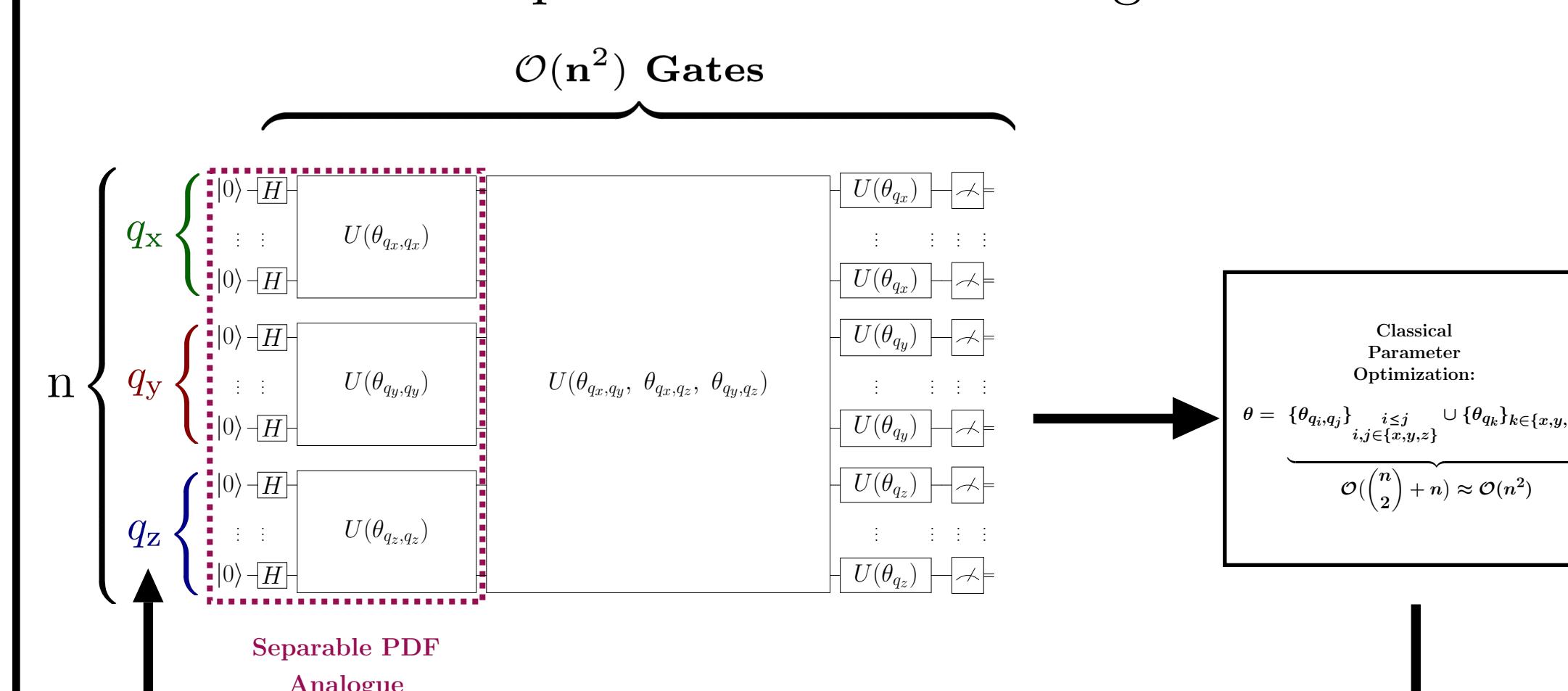
2. The Quantum Algorithm

I) Discretise the multidimensional integration domain



$f(x,y,z)$

II) Map the grid to a PQC. Quantum Generative Modelling, in particular a Quantum Circuit Born Machine (QCBM), adapts the PDF over the grid

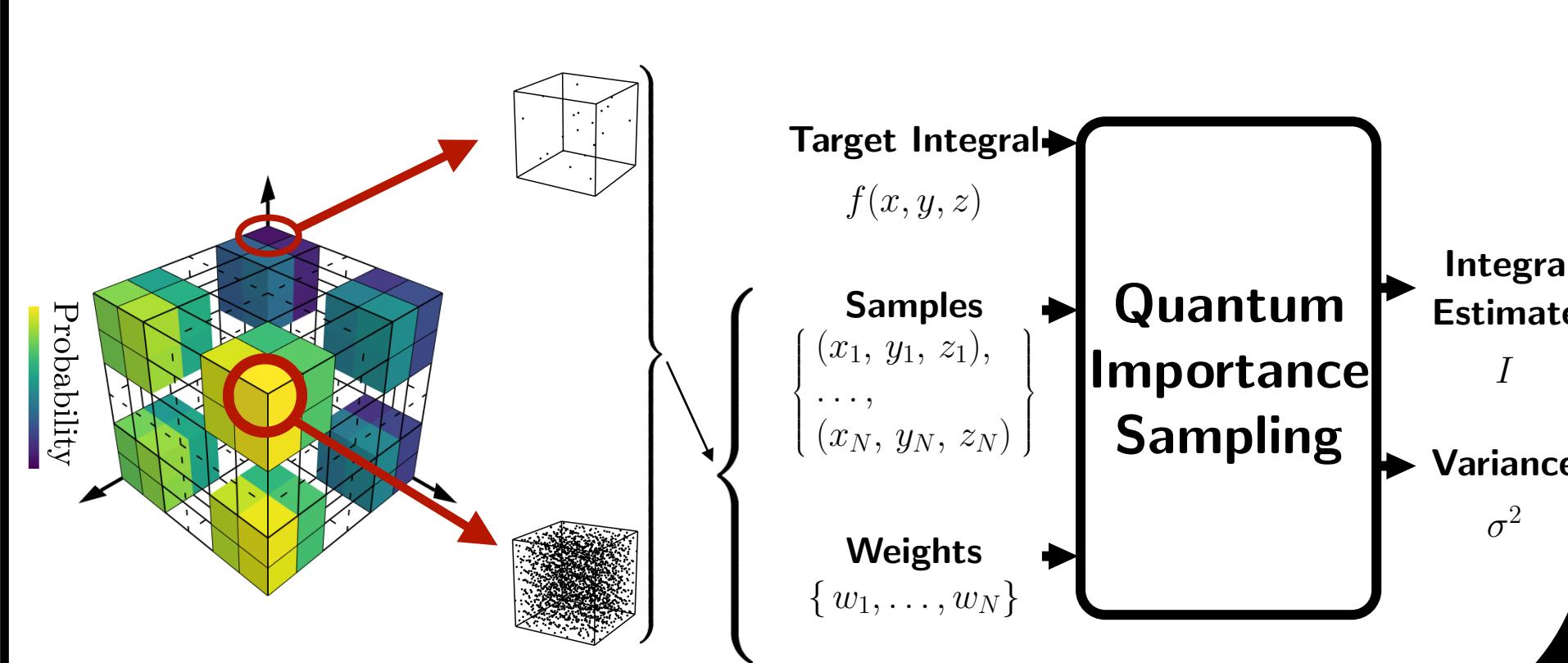


III) Sample from the optimised PQC.

Shot counts per state set

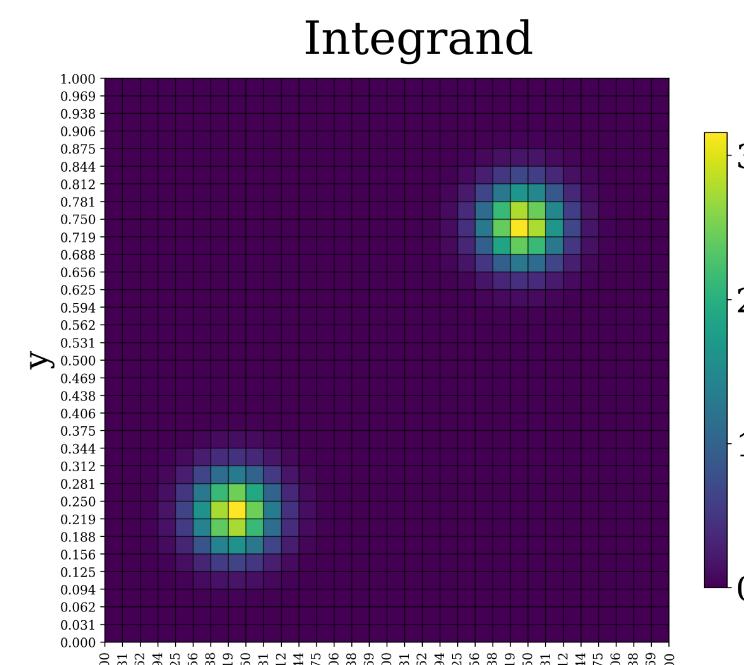
the quasi-random points allocation.

Samples and weights produce the estimation.

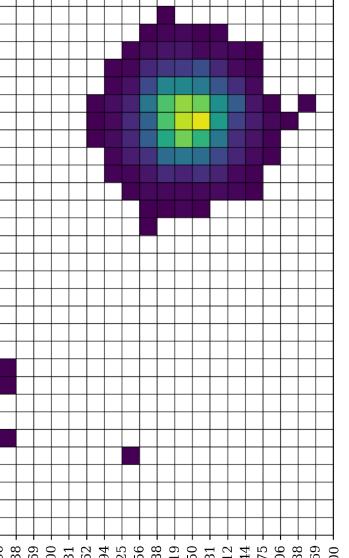


3. Two-Dimensional Gaussian Case

$$\int_{[0,0]}^{[1,1]} f(\mathbf{x}) d\mathbf{x} = \int_{[0,0]}^{[1,1]} \left(\sum_{i=0}^1 e^{-200|\mathbf{x}-\mathbf{r}_i|^2} \right) d\mathbf{x}, \quad \mathbf{r}_0 = (0.23, 0.23), \quad \mathbf{r}_1 = (0.74, 0.74)$$

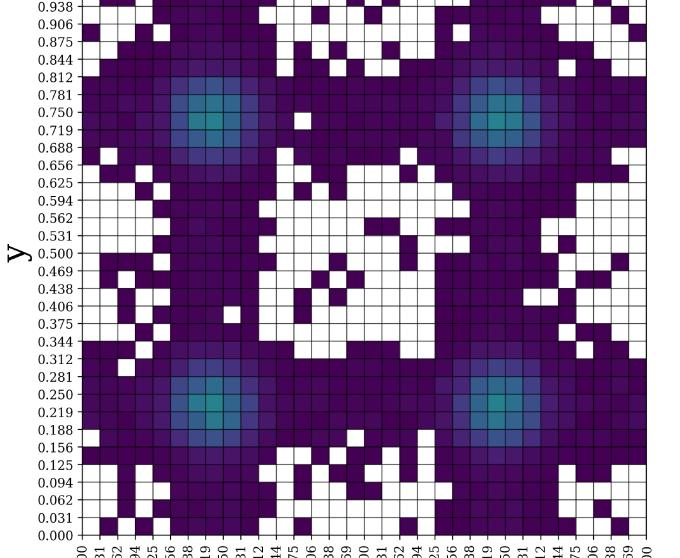


Integrand



QAIS

VEGAS in the QAIS Grid



$$\text{Naive QAIS integral estimator: } \hat{I}_N^{(\text{QAIS})} = \frac{1}{N} \sum_{j=1}^N w^{(i)} f(\mathbf{x}_j^{(i)}), \quad w^{(i)} = \frac{|\Omega^{(i)}| N}{N_i}.$$

$$(\hat{\sigma}_N^{(\text{QAIS})})^2 = \frac{1}{N-1} \left(\frac{1}{N} \sum_{j=1}^N \left[w^{(i)} f(\mathbf{x}_j^{(i)}) \right]^2 - (\hat{I}_N^{(\text{QAIS})})^2 \right)$$

$\Omega^{(i)}$: i -th grid cell $\equiv i$ -th quantum state
 $\mathbf{x}_j^{(i)}$: j -th random point at grid cell i
 N : Number of shots
 N_i : Number of times state i was measured

4. Debiasing Through Tiling Algorithm

1. Challenge: Unmeasured states \rightarrow Integral underestimation (Bias)

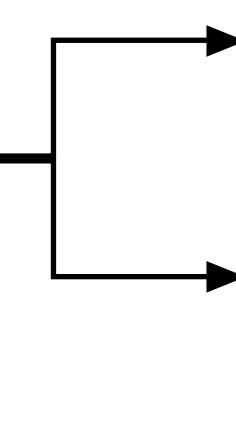
Efficient Quantum Measurement:

Target at few states

Unbiasedness:

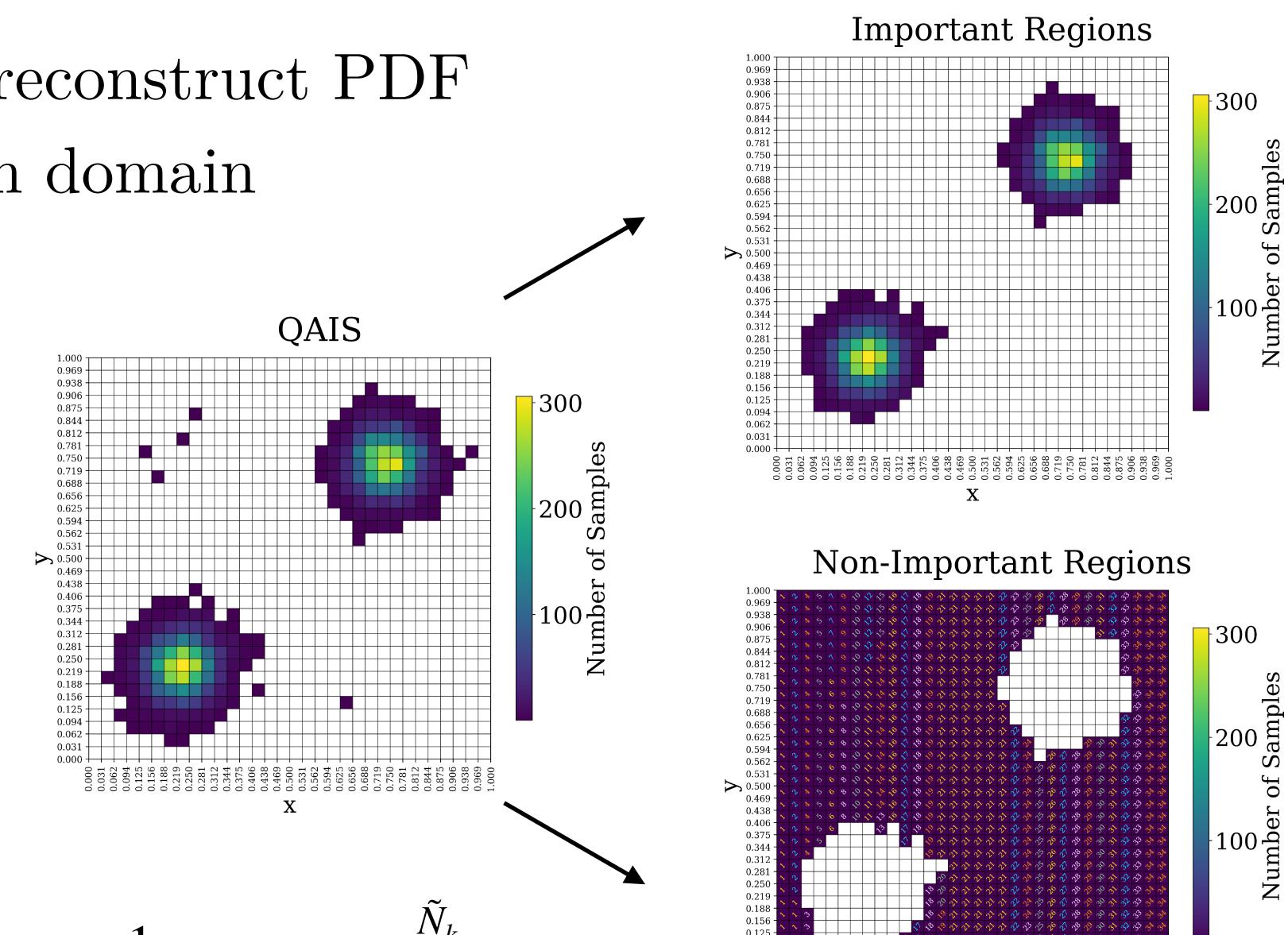
Evaluate every grid cell

2. Conflicting objectives:



3. Solution: Tiling Algorithm to reconstruct PDF over the full integration domain

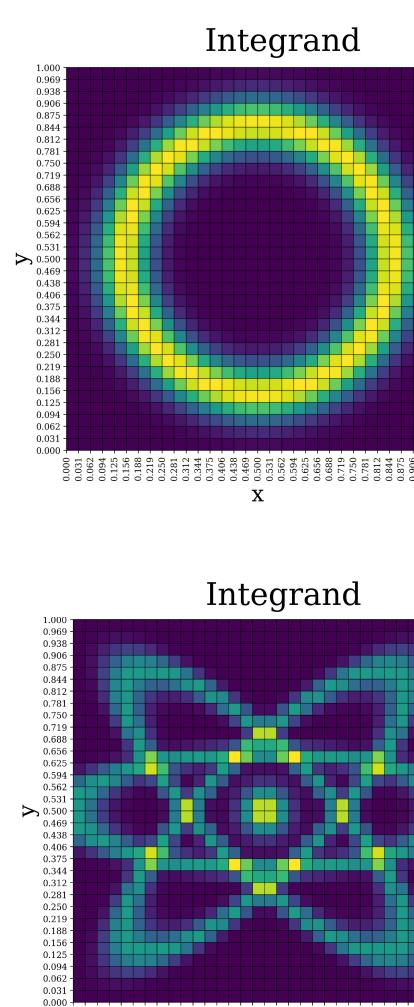
I) Detailed sampling in Important Regions



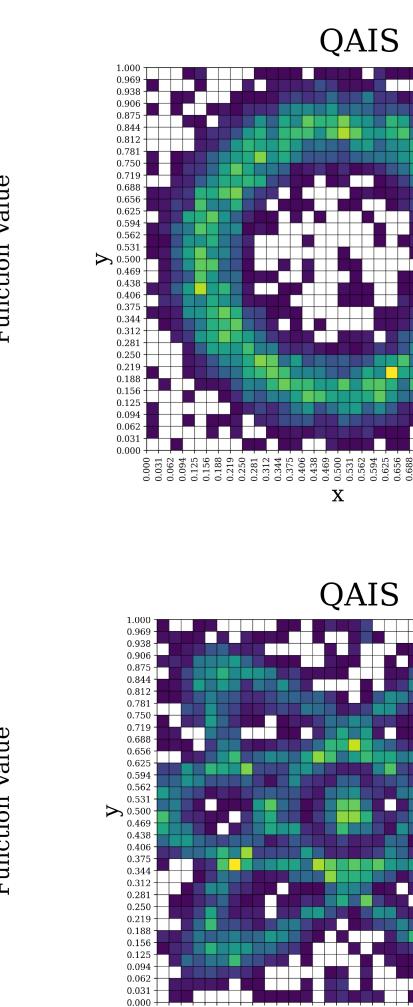
II) Sparse uniform sampling in the Non-Important Region

$$\text{QAIS estimator: } \hat{I}_N^{(\text{QAIS})} = \frac{1}{N} \sum_{i \in \Omega_I}^{N_I} \sum_{j=1}^{N_i} f(\mathbf{x}_j^{(i)}) + \frac{1}{N} \sum_{k \in \Omega_{N-I}}^{\tilde{N}_k} \sum_{j=1}^{\tilde{N}_k} f(\mathbf{x}_j^{(k)})$$

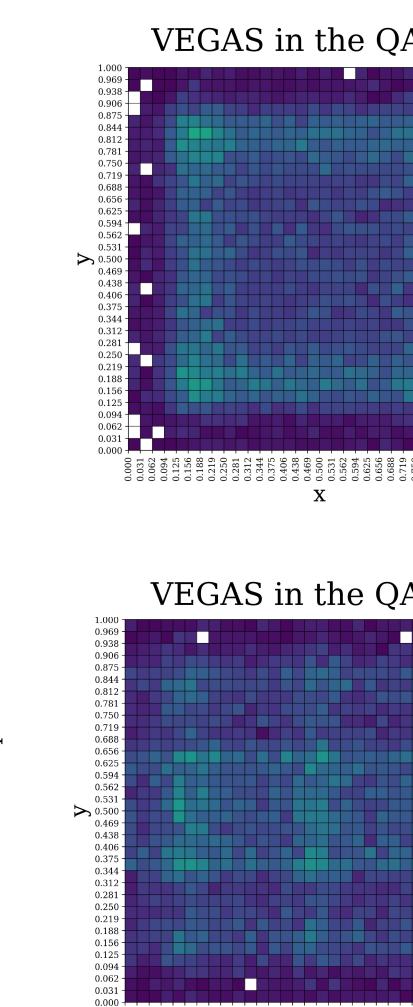
5. Non-Trivial Two-Dimensional Cases



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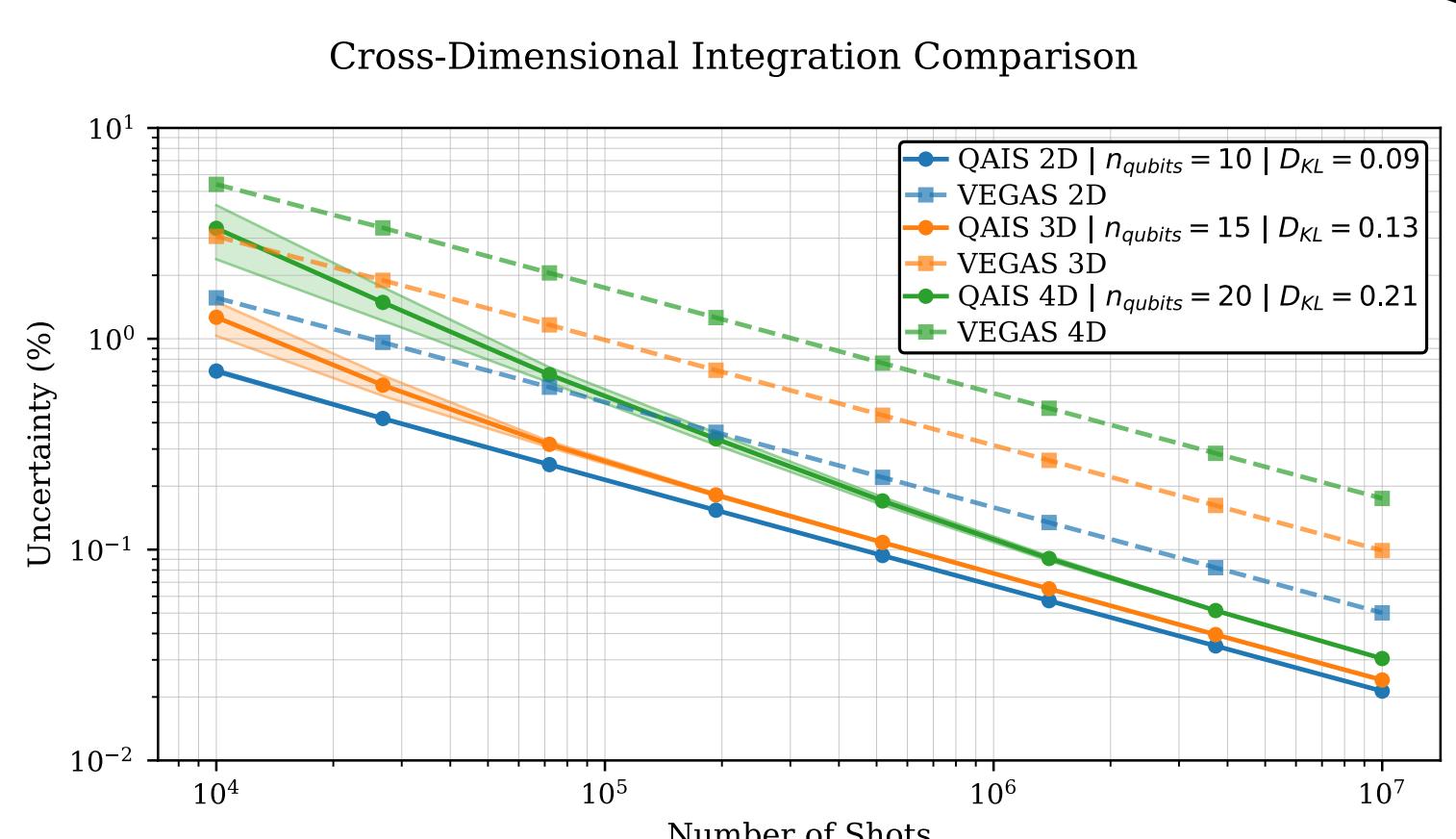
VEGAS in the QAIS grid

6. Cross-Dimensional Results

Dimensions: 2-4

$$\int_{[0,1]^d} f(\mathbf{x}) d\mathbf{x} = \int_{[0,1]^d} \left(\sum_{i=0}^2 e^{-50|\mathbf{x}-\mathbf{r}_i|^2} \right) d\mathbf{x}$$

$\mathbf{r}_0 = (0.23, \dots, 0.23)$,
 $\mathbf{r}_1 = (0.39, \dots, 0.39)$,
 $\mathbf{r}_2 = (0.74, \dots, 0.74)$



7. Summary

- Entangled PQC: Encodes non-separable proposal PDFs that can capture the intricate structures of integrands.
- High-precision: Integral estimates stay robust even as the dimensionality increases.
- Current bottleneck: Optimisation through QCBM is significantly more expensive than VEGAS.
- Future work: Loading the proposal PDFs more efficiently.

References

- [1] G. P. Lepage, A New Algorithm for Adaptive Multidimensional Integration, *J. Comput. Phys.* 27 (1978) 192.
- [2] K. Pyretzidis, J. J. Martínez de Lejarza and G. Rodrigo, Unlocking Multi-Dimensional Integration with Quantum Adaptive Importance Sampling, arXiv:2506.19965 [quant-ph] (2025).



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