

# Variational Quantum Solutions to the Shortest Vector Problem

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## Abstract

We explore how (efficiently) Noisy Intermediate Scale Quantum (NISQ) devices may be used to solve SVP by mapping the problem to that of finding the ground state of a suitable Hamiltonian. In particular, (i) we propose an approach to reduce the number of required qubits to  $\approx 10^3$  to tackle instances on the edge of classical capabilities; (ii) we exclude the zero vector from the optimization space by proposing (a) a different classical optimisation loop or alternatively (b) a different mapping to the Hamiltonian. Full paper [1].

## Shortest Vector Problem

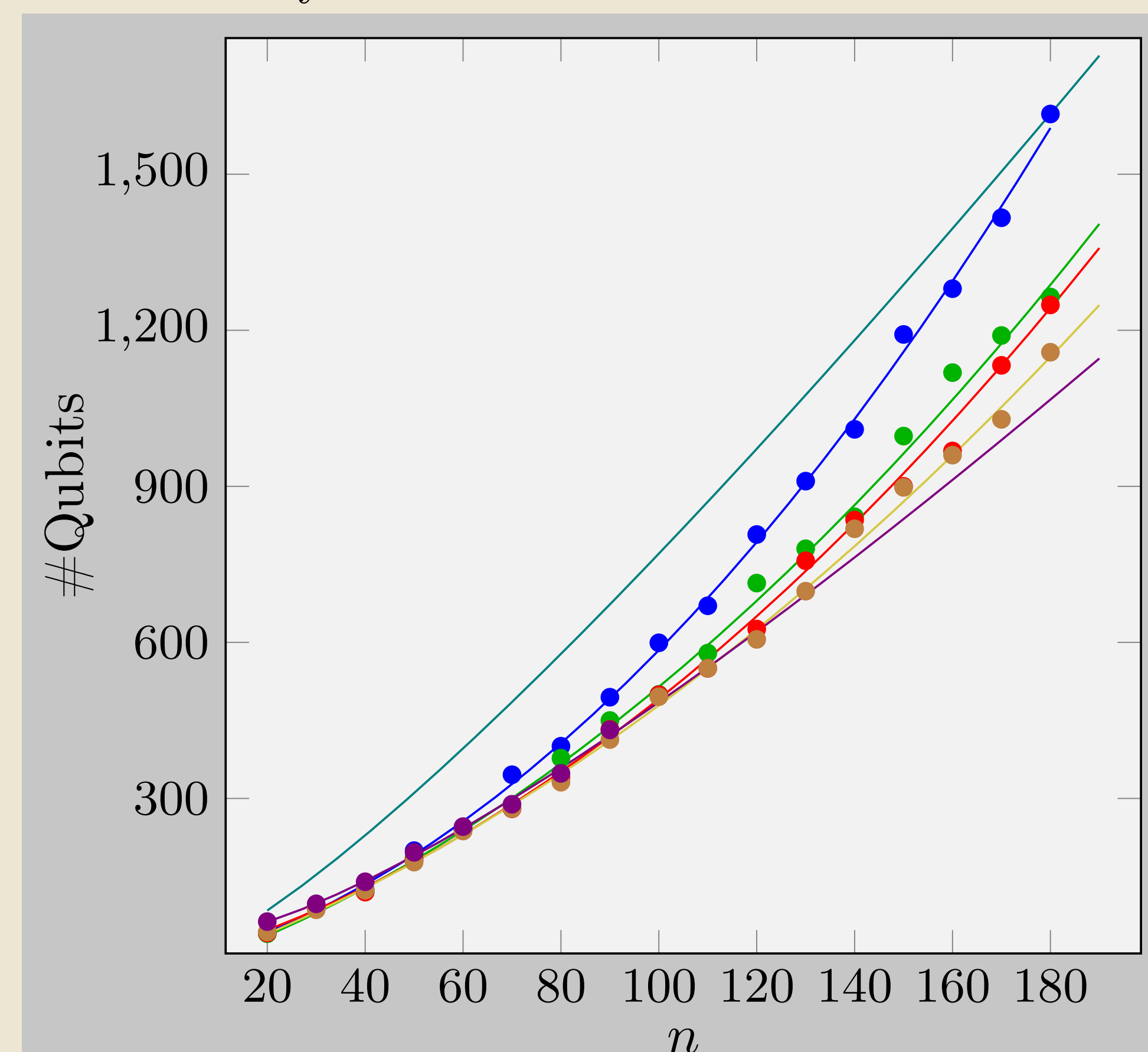
Given an integer lattice basis  $B$ , the SVP finds the shortest non-zero vector of lattice  $\mathcal{L}(B) = \{Bx : x \in \mathbb{Z}^n\}$  denoted by  $\lambda(\mathcal{L}) = \min\{\|y\|_p : y \in \mathcal{L}, y \neq \mathbf{0}\}$ . NP-Hardness of SVP has been shown for  $p = \infty$  [3] and for  $p = 2$  under randomized reductions [4]. Although not proven, hardness of SVP is also conjectured in quantum settings. It is particularly appealing to cryptography as **many quantum-safe classical cryptographic protocol proposals are based on the hardness of SVP.**

## Variational Q. Algorithms

Variational Quantum Algorithms are promising candidates for NISQ era due to low qubit requirements and partial resilience against noise without quantum error correction. Given a problem encoded as ground state of Hamiltonian  $\mathcal{H}$ , they utilize classical optimization to find  $\theta$  minimizing a cost  $C(\theta) = \min_{\theta} \langle \psi(\theta) | \mathcal{H} | \psi(\theta) \rangle$  evaluated on a quantum device. There exists a natural mapping of **Quadratic Unconstrained Binary Optimization (QUBO)** problem formulation to Ising Hamiltonians.

## Estimated Qubit Scaling

Average qubit requirements to encode SVP in a problem Hamiltonian.  $n = 180$  is an upper bound on capability of classical SVP solvers.



Basis preprocessing: LLL, BKZ-20, BKZ-50, BKZ-70, pseudo-HKZ[1], q.enum HKZ[1]

## Mapping SVP to a Hamiltonian Operator

Given an  $n$ -dimensional full-rank row-major lattice basis matrix  $B$ , let  $G = BB^T$ . The shortest non-zero lattice vector can be found by solving the following **integer constrained optimization problem**:

$$[\lambda(\mathcal{L})]^2 = \min_{y \in \mathcal{L}(B) \setminus \{0\}} \|y\|^2 = \min_{x \in \mathbb{Z}^n \setminus \{0\}} \sum_{i=1}^n x_i G_{ii} + 2 \sum_{1 \leq i < j \leq n} x_i x_j G_{ij}.$$

To construct a **QUBO** formulation we propose the following:

### 1. Conversion to a binary optimization problem

To express  $x_i$  as a finite sum of binary variables, bounds  $|x_i| \leq a_i$  that are **sufficient** (encode the SVP solution) and **efficient** (realistic qubit overhead) need to be determined. Letting  $\hat{B} := (BB^T)^{-1}B$  be a specific basis of a dual lattice  $\mathcal{L}^* = \{y \in \mathbb{R}^n : \forall x \in \mathcal{L} \langle x, y \rangle \in \mathbb{Z}\}$ , the following results improve the estimates on qubit requirements for solving the SVP with VQAs. Assuming a bound  $A$  on the SVP solution is known a priori (e.g. Gaussian Heuristic) we can bound each individual element of  $x$ :

**Lemma [1].** Let  $x_1, \dots, x_n$  be such that  $\|x_1 \cdot \vec{b}_1 + \dots + x_n \cdot \vec{b}_n\| \leq A$ , then for all  $i = 1, \dots, n$  we have  $|x_i| \leq A \|\vec{b}_i\|$  where  $\vec{b}_1, \dots, \vec{b}_n$  are the rows of  $\hat{B}$  and  $B$  is the matrix whose rows are  $\vec{b}_1, \dots, \vec{b}_n$ .

This allows for asymptotic estimation of qubit scaling with  $\delta(\hat{B}) = 2^{\mathcal{O}(n^2)}$  being orthogonality defect<sup>1</sup>:

**Corollary [1].** The number of qubits required for the enumeration on the basis  $B$ , assuming the Gaussian heuristic with multiplicative factor  $C$ , is bounded by  $2n + \log_2 \left( \left( \frac{C^2 n}{2\pi e} \right)^{n/2} \delta(\hat{B}) \right)$ .

### 2. Avoiding the constraint by optimizing towards the 1st excited state

We have analyzed two possibilities that differ by suitable quantum computational models:

- **Modify the cost function**  $C'(\theta) := \frac{1}{1 - |\langle \psi(\theta) | \psi_0 \rangle|^2} \langle \psi(\theta) | H | \psi(\theta) \rangle$  to penalize states proportionally to their overlap with the ground state. The method does not increase qubit requirements, but due to classical cost post-processing, is suitable only for *Variational Quantum Eigensolver (VQE)* algorithm.
- **Construct a new Ising Hamiltonian by encoding a penalty term.** This approach is suitable if SVP is to be tackled by *Quantum Approximate Optimization Algorithm, Adiabatic Quantum Computation* or *Quantum Annealing*.  $n$  additional binary variables  $\{\zeta_i\}_{i=1, \dots, n}$  are to be introduced with bijective correspondence to  $\{x_i\}_{i=1, \dots, n}$ .

If the bound  $|x_i| \leq a$  is determined then  $x_i$  can be encoded as

$$x_i = -a + \zeta_i a + \omega_i (a + 1) + \sum_{j=0}^{\lfloor \log(a-1) \rfloor - 1} 2^j \tilde{x}_{ij} + (a - 2^{\lfloor \log(a-1) \rfloor}) \tilde{x}_{i, \lfloor \log(a-1) \rfloor} \quad (1)$$

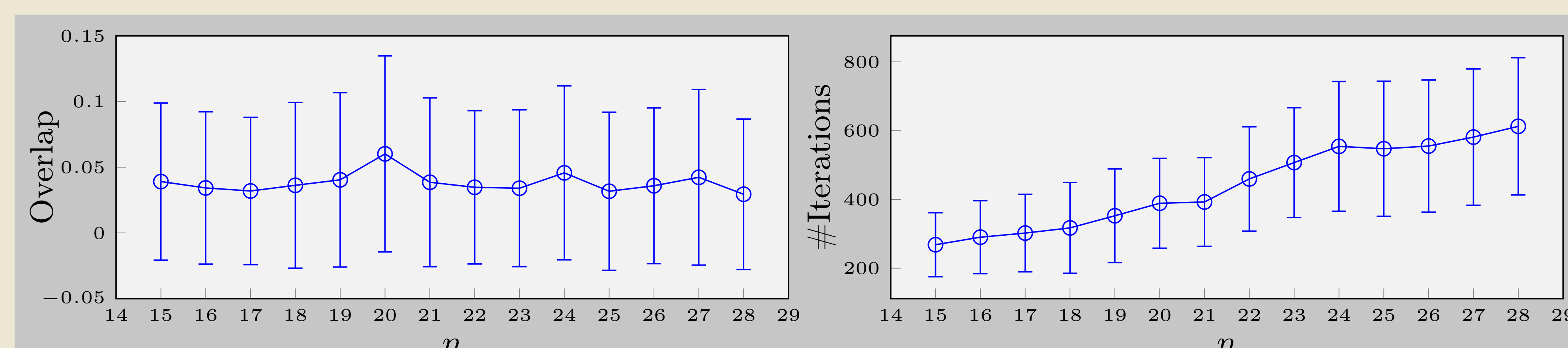
It follows that  $x_i = 0 \implies \zeta_i = 1$  and the penalization term  $L \prod \zeta_i$  (expressed as a QUBO term below) introduces penalty  $L \gg 0$  iff  $\forall \zeta_i = 1$ .

$$L \prod \zeta_i = L \left( 1 + \sum_{i=1}^n z_i (-1 - \zeta_i) + \sum_{k=i+1}^n (1 - \zeta_k) \right) \quad (2)$$

<sup>1</sup>True for LLL or BKZ reduced basis.  $\delta(\hat{B}) = 2^{\mathcal{O}(n \log(n))}$  if basis is quasi-HKZ [1] reduced.

## Classical Emulation of the Quantum SVP Approach

SVP approached by **VQE** was emulated using *FastVQA*[2] library omitting effects of noise up to 28 dimensions of qary lattice instances, setting a new record in the existing literature [1]. Constant overlap  $\langle \psi(\theta_{\text{returned by VQE}}) | \text{ground\_state}(\mathcal{H}) \rangle \approx 4\%$  and linear time scaling have been observed.



## References

- [1] M. R. Albrecht, M. Prokop, Y. Shen, and P. Wallden, Variational quantum solutions to the Shortest Vector Problem. arXiv, 2022. doi: 10.48550/ARXIV.2202.06757.
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- [4] M. Ajtai, The shortest vector problem in L2 is NP-hard for randomized reduction, in "Proc. 30th ACM Symposium on Theory of Computing (STOC), 1998."