

Track #2 - Real-World Optimization Problem - Knapsack

Shrinidhi Mahesh, Grace Chowdhry, Asmita Mohanty, Jai Veilleux
Superposition Squad

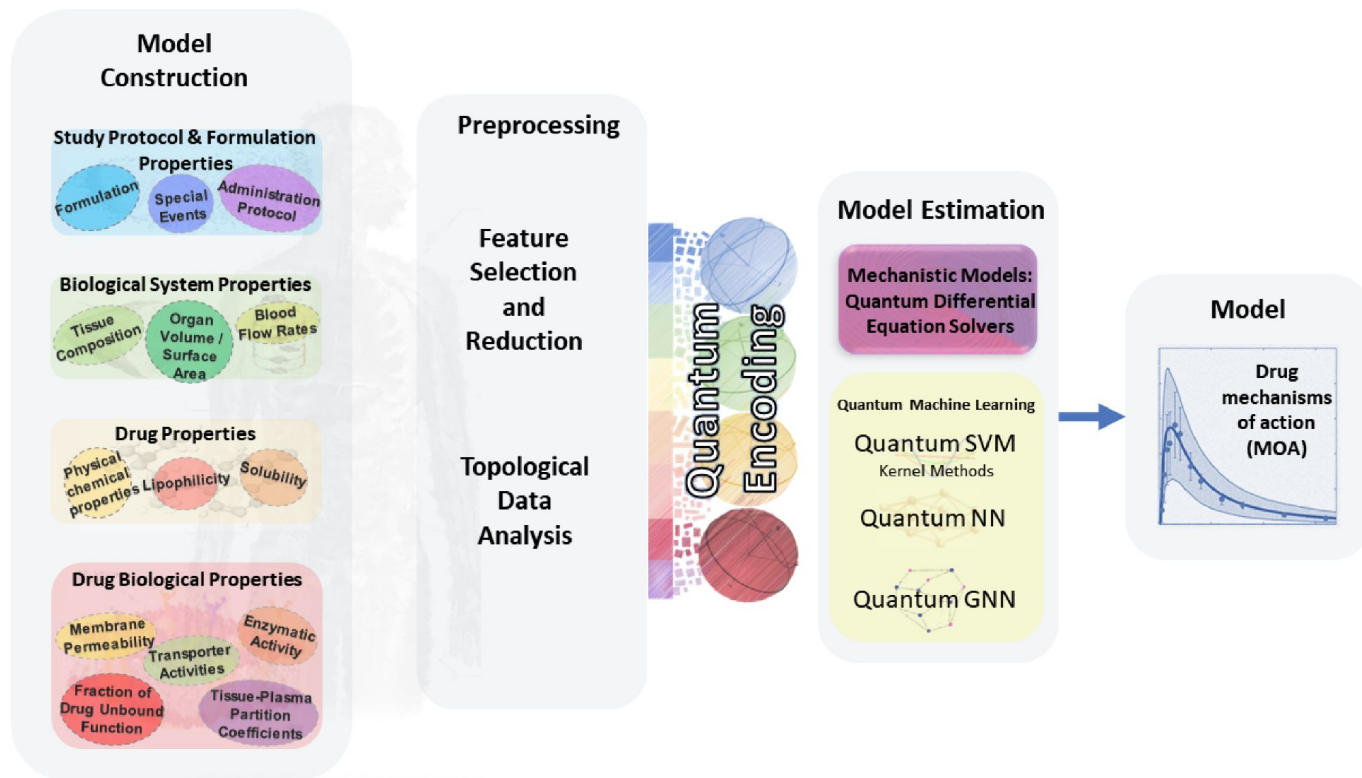
What is the *knapsack* problem?

- A combinatorial optimization problem that states :

Given a set of items, each with a weight and a value, determine which items to include in the collection so that the total weight is less than or equal to a given limit and the total value is as large as possible.

- Often arises in resource allocation where decision-makers have to choose from a set of non-divisible projects or tasks under a fixed budget or time constraint, respectively.
- This decision problem form of the knapsack problem is NP-complete which means that there is no known algorithm that is both correct and fast (polynomial-time) in all cases making it the absolute best problem to solve using quantum computing.

Quantum Knapsack in Real World Drug Discovery



Real World Constraints in Implementation

- Technical Constraints
 - Scalability issues due to hardware limitation.
 - High qubit use from pairwise connections
 - Highly sensitive to penalty terms
- Practical Concerns
 - Upfront costs are high
 - If candidate drugs are selected by underdeveloped quantum–classical hybrid pipelines, explaining decision-making for regulatory approval could be difficult.
- Governance
 - Possibility of misusing tech to create bioweapons and drugs - need robust policies to counter
 - Due to cost & scarcity of hardware only well-funded institutions may benefit, widening gaps in drug-discovery innovation possibly creating monopolies

The Hamiltonian

$$H(x) = - \sum_i v_i x_i + A(\sum_i w_i x_i - C)^2$$

$$x_i = \frac{1-Z_i}{2}$$

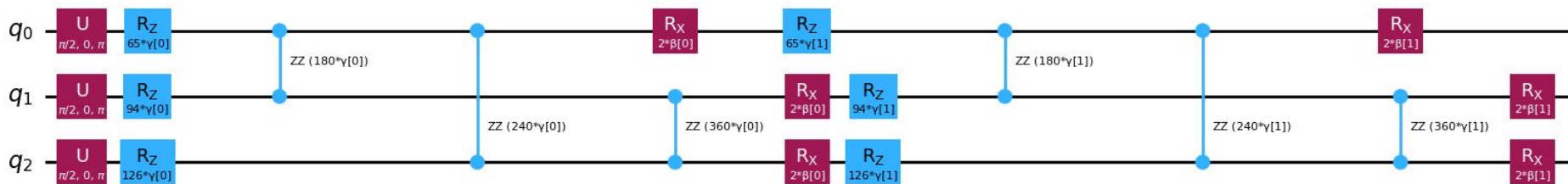
$$v_i x_i = \frac{v_i}{2} I - \frac{v_i}{2} Z_i$$

$$w_{ij} x_i x_j = \frac{w_{ij}}{4} I - \frac{w_{ij}}{4} Z_i - \frac{w_{ij}}{4} Z_j + \frac{w_{ij}}{4} Z_i Z_j$$

- x_i represents a binary select / do not select (1, 0)
 - Map to (1, -1) as orthogonal states are distinguishable
- A is a penalty term, and generally should be $\geq \max(w)$
- C is the capacity of the knapsack
- v_i and w_i are value and weight

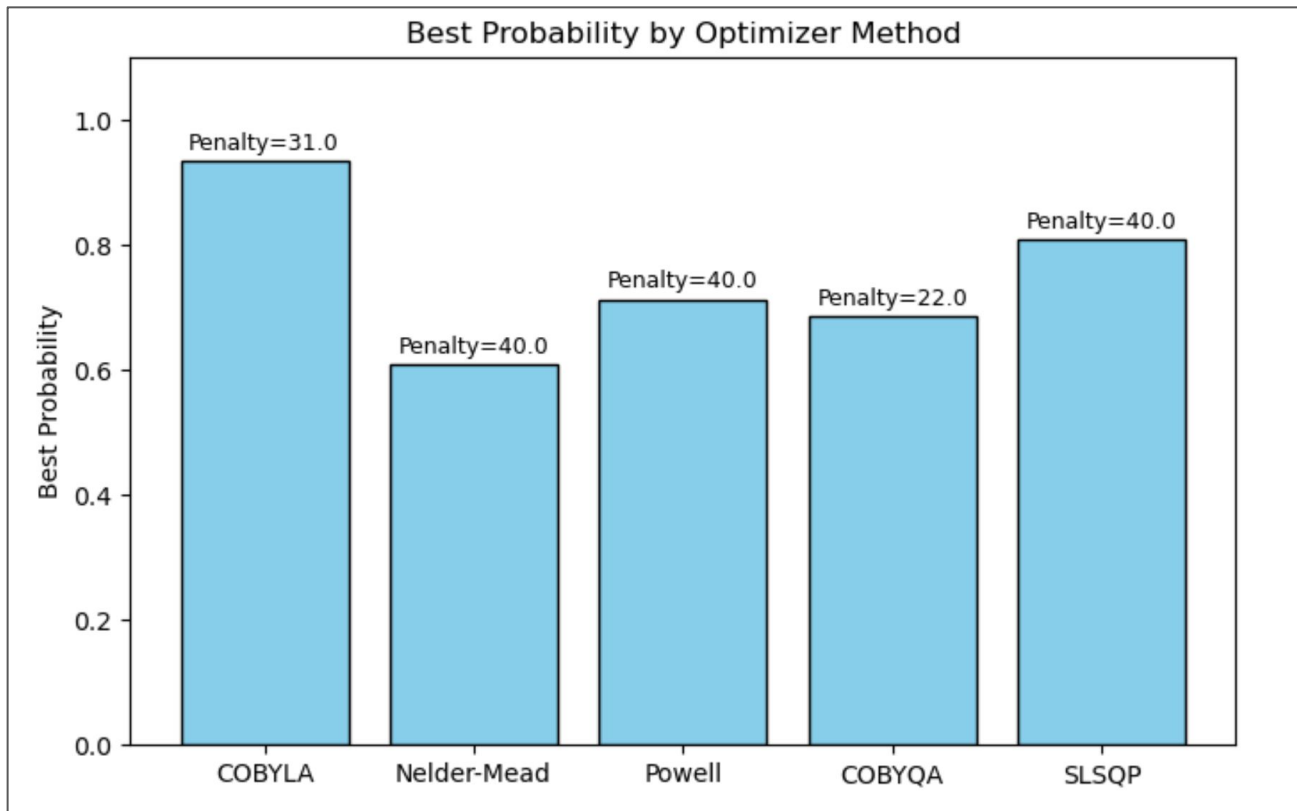
The QAOA Algorithm

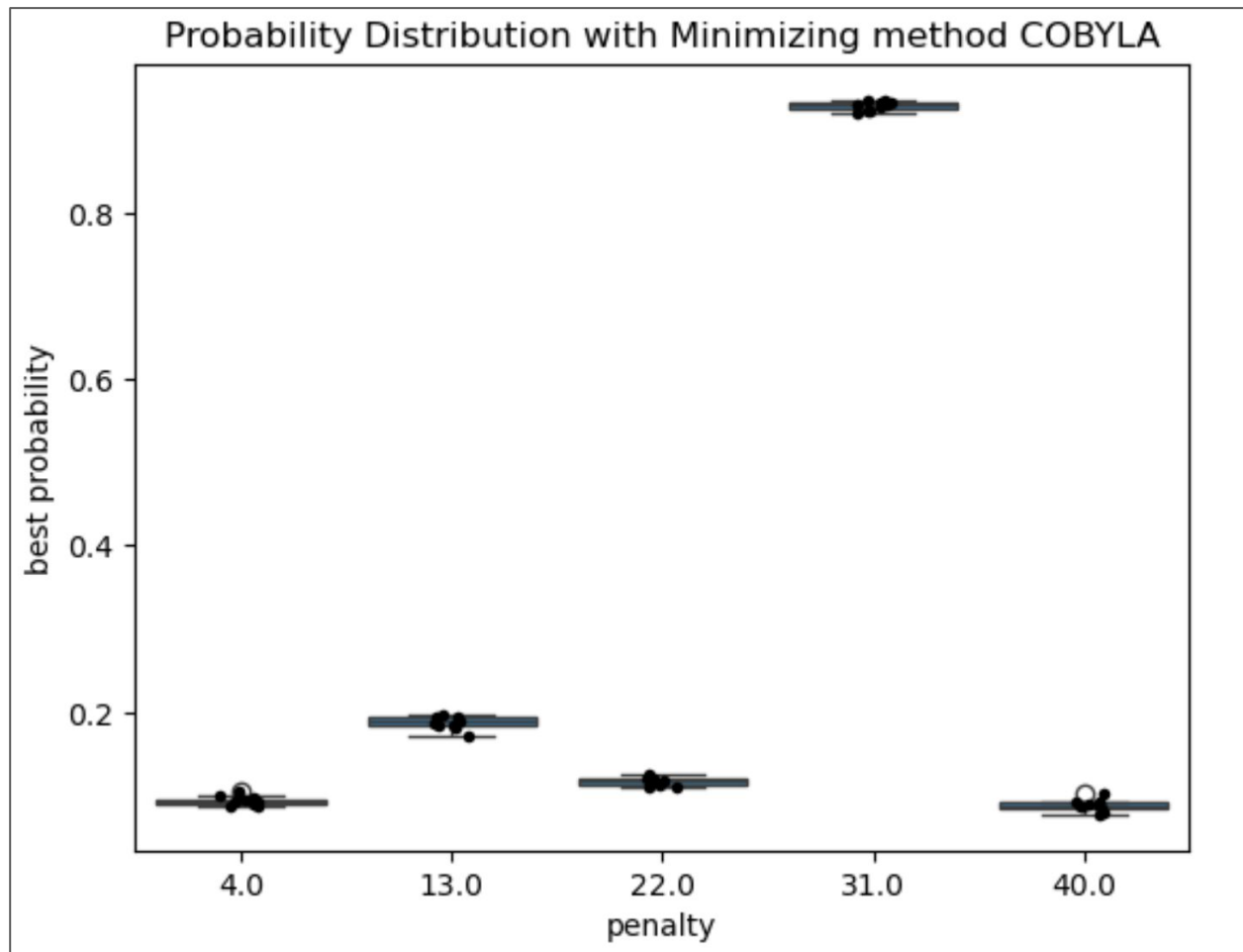
- We are minimizing the expectation value
 - $\langle \psi(\gamma, \beta) | H | \psi(\gamma, \beta) \rangle$
 - A classical optimizer tunes γ and β to reach this minimum
- Once converged, sampling the tuned circuit should produce the bitstring representing the max items with a higher percentage than other options in the search space



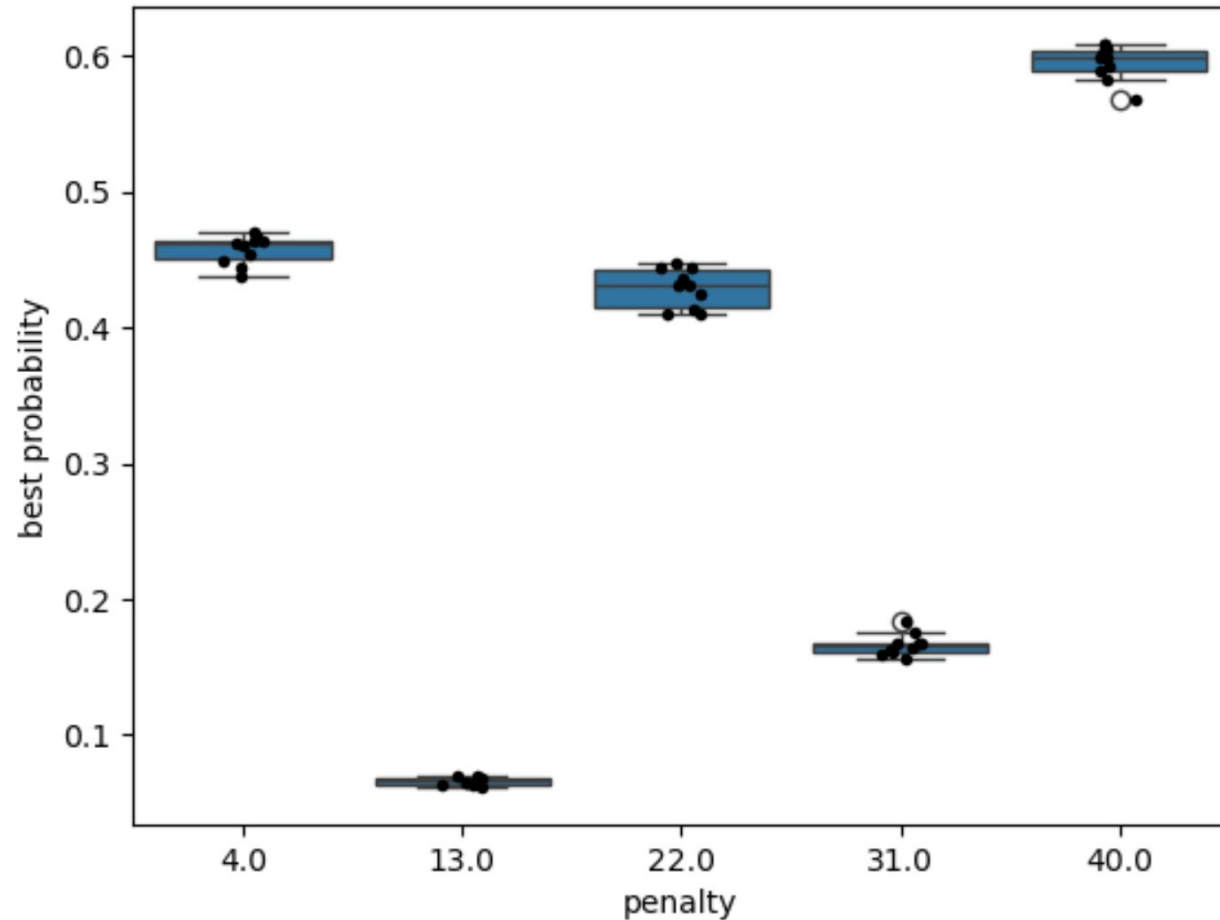
QAOA ansatz with 2 repetitions of the Hamiltonian

Fine Tuning the Minimization Algorithm

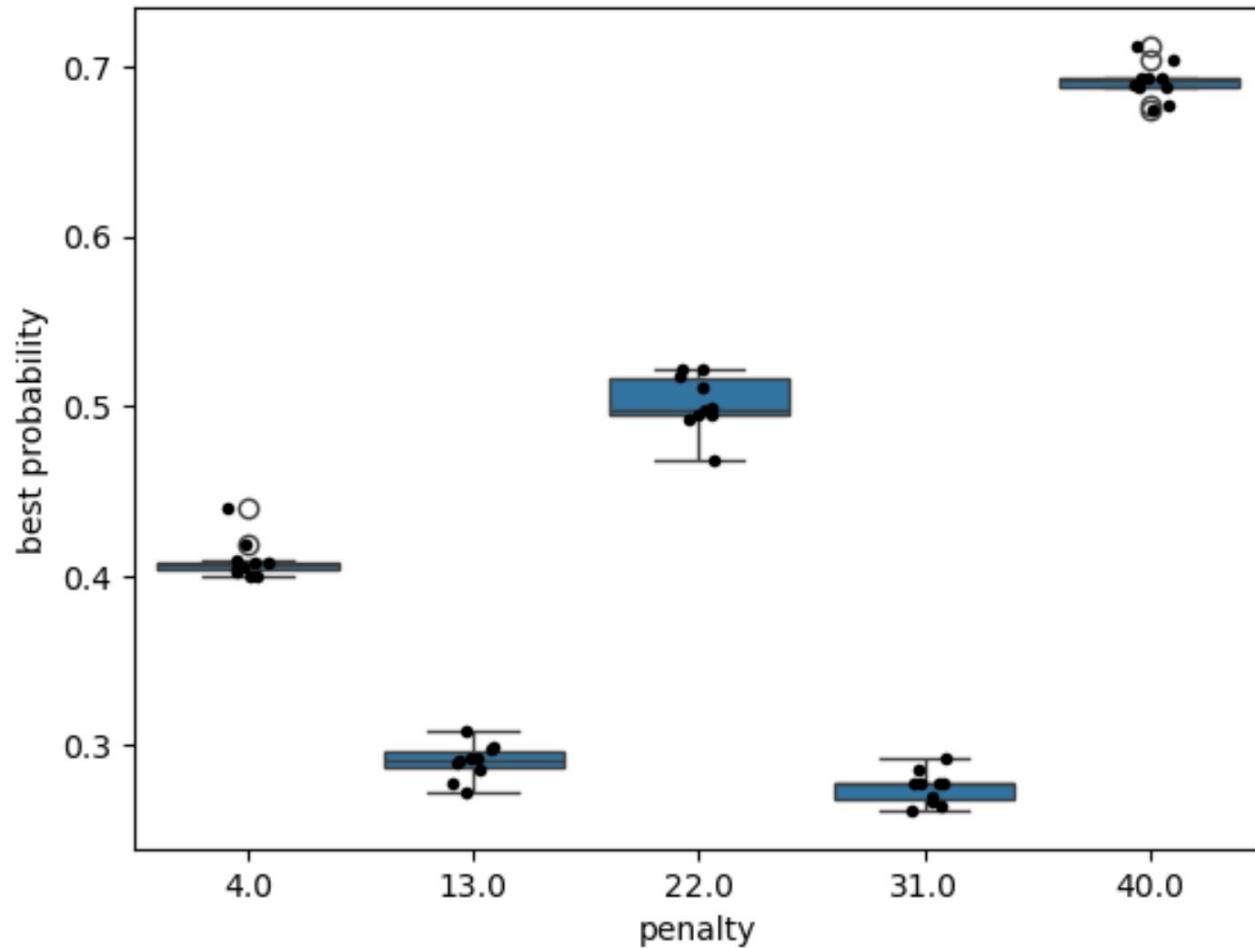




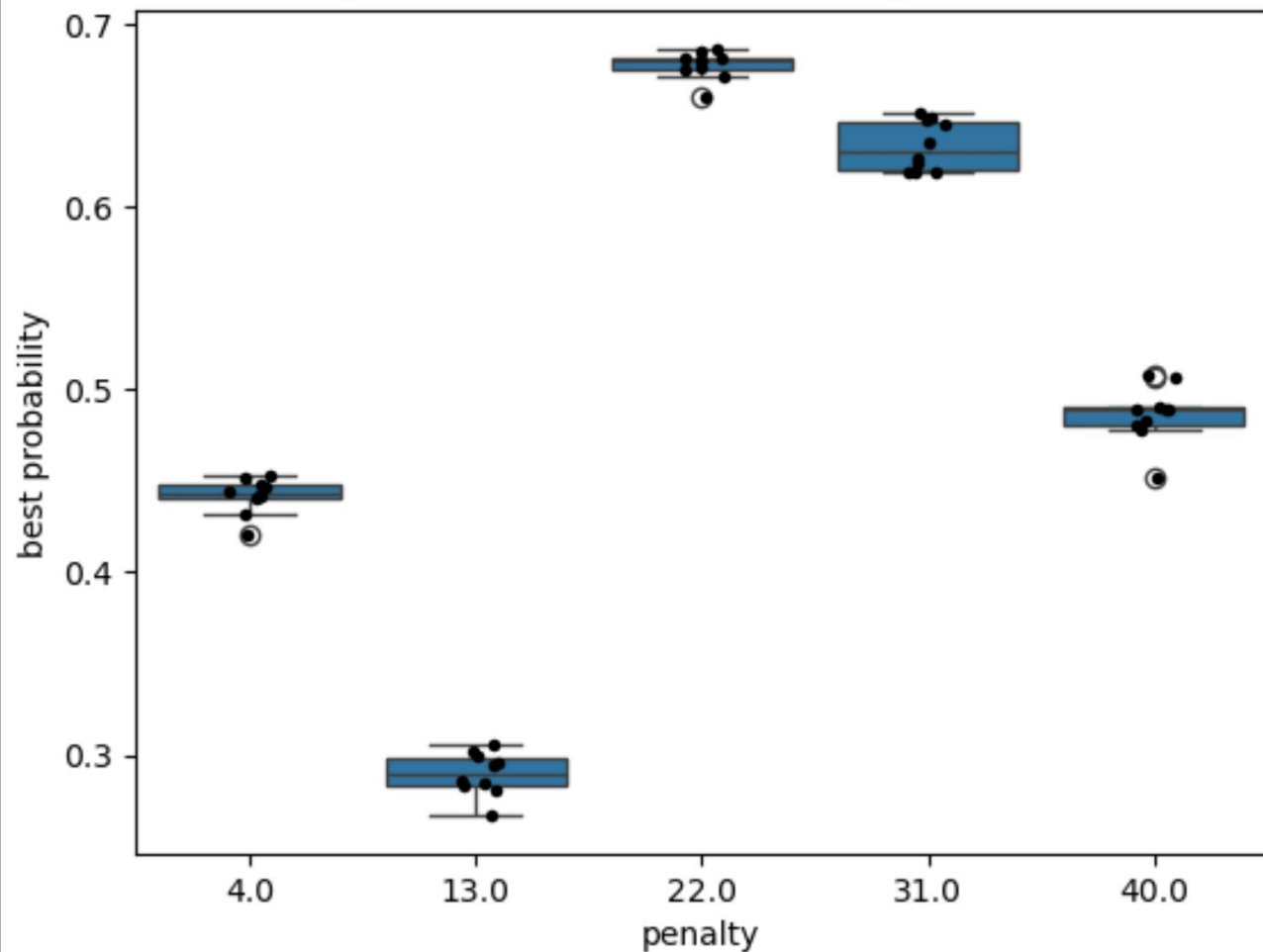
Probability Distribution with Minimizing method Nelder-Mead



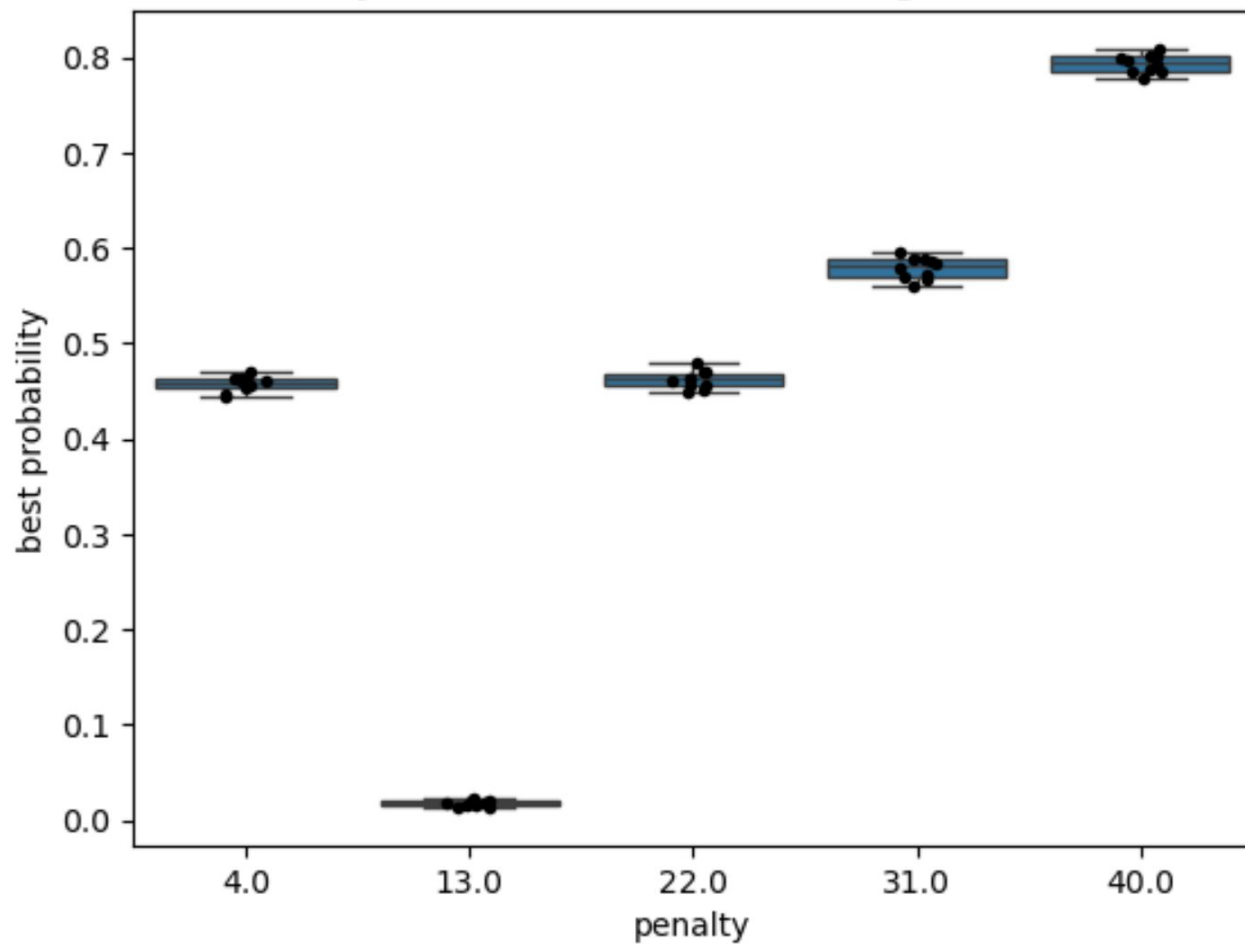
Probability Distribution with Minimizing method Powell



Probability Distribution with Minimizing method COBYQA



Probability Distribution with Minimizing method SLSQP



Thank You!

Citations

How Quantum Computing Can Be Used for Drug Development. (2022, July 15). Classiq.io; Classiq Technologies.

<https://www.classiq.io/insights/quantum-computing-for-drug-development>

Knapsack problem. (2020, September 5). Wikipedia. https://en.wikipedia.org/wiki/Knapsack_problem

Rawal, B., & Braga, D. M. (2024). Harnessing AI and Quantum Computing: Revolutionizing Drug Discovery and Approval Processes Using Collagen as an Example (Preprint). *JMIR Bioinformatics and Biotechnology*.

<https://doi.org/10.2196/69800>

Towards quantum computing for clinical trial design and optimization: A perspective on new opportunities and challenges. (2016). Arxiv.org. <https://arxiv.org/html/2404.13113v1#S4>