

Parameterized Quantum Circuits and Bayesian Optimization for violating the Bell Inequality

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Introduction

Aim

Win a formulation of the CHSH game

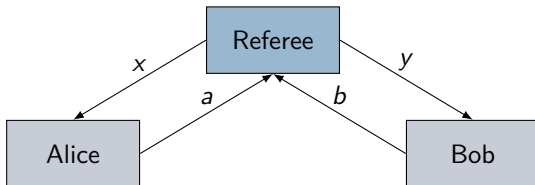
Methodology

Training a simple parameterized quantum circuit

Data

Outputs of quantum circuits simulations

Problem Formulation: CHSH Game [1]



- 1 The referee sends a bit x to Alice and a bit y to Bob. These bits are drawn from two Bernoulli-distributed random variables:

$$X \sim \mathcal{B}(0.5), \quad Y \sim \mathcal{B}(0.5)$$

- 2 Alice and Bob send back, respectively, bits a and b .

Alice and Bob **win the game** if¹:

$$x \wedge y = a \oplus b$$

(1)

¹Where \wedge denotes the logical *AND* and \oplus the logical *XOR* operations.

Strategy [1]

Naming $p_{AB|XY}(a, b|x, y)$ the conditional distribution which relates to the **strategy** operated by Alice and Bob, we define the winning function $V : \{0, 1\}^4 \rightarrow \{0, 1\}$ as:

$$V(x, y, a, b) := \begin{cases} 1 & \text{if } x \wedge y = a \oplus b, \\ 0 & \text{otherwise.} \end{cases} \quad (2)$$

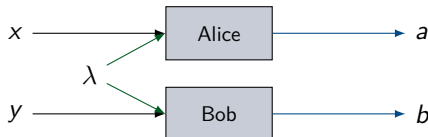
and, hence, express the probability of winning the CHSH game as

$$\begin{aligned} \Pr\{\text{win}\} &= \sum_{a,b,x,y} V(x, y, a, b) p_{AB|XY}(a, b|x, y) p_{XY}(x, y) = \\ &= \frac{1}{4} \sum_{a,b,x,y} V(x, y, a, b) p_{AB|XY}(a, b|x, y) \end{aligned} \quad (3)$$

Assumption: there is a random variable Λ that describes the strategy:

$$p_{AB|XY}(a, b|x, y) = \int d\lambda p_{AB|\Lambda XY}(a, b|\lambda, x, y) p_{\Lambda|XY}(\lambda|x, y) \quad (4)$$

Classical Strategy [1]



We have conditional independence:

Also it is always possible to write:

$$\begin{cases} p_{\Lambda|XY} = p_{\Lambda}(\lambda) \\ p_{AB|\Lambda XY} = p_{A|\Lambda X}(a|\lambda, x) p_{B|\Lambda Y}(b|\lambda, y) \end{cases} \quad \begin{cases} p_{A|\Lambda X}(a|\lambda, x) = \int dn f(a|\lambda, x, n) p_N(n) \\ p_{B|\Lambda Y}(b|\lambda, y) = \int dm g(b|\lambda, y, m) p_M(m) \end{cases}$$

$$\begin{aligned} p_{AB|XY} &= \int \int \int d\lambda dn dm f(a|\lambda, x, n) g(b|\lambda, y, m) p_{\Lambda}(\lambda) p_N(n) p_M(m) \\ &= \int d\lambda f'(a|\lambda, x) g'(b|\lambda, y) p_{\Lambda}(\lambda) \end{aligned} \quad (5)$$

where f, g, f' and g' are deterministic (binary-valued) functions.

Classical Strategy - II [1]

We can then write

$$\begin{aligned}
 \Pr\{\text{win}\} &= \frac{1}{4} \sum_{a,b,x,y} V(x,y,a,b) p_{AB|XY}(a,b|x,y) \\
 &= \frac{1}{4} \sum_{a,b,x,y} V(x,y,a,b) \int d\lambda f'(a|\lambda,x) g'(b|\lambda,y) p_{\Lambda}(\lambda) \\
 &= \int d\lambda p_{\Lambda}(\lambda) \left[\frac{1}{4} \sum_{a,b,x,y} V(x,y,a,b) f'(a|\lambda,x) g'(b|\lambda,y) \right] \\
 &\leq \frac{1}{4} \sum_{a,b,x,y} V(x,y,a,b) f'(a|\lambda^*,x) g'(b|\lambda^*,y) \tag{6}
 \end{aligned}$$

and conclude that deterministic strategies are optimal among all classical strategies.

Deterministic Classical Strategy [1]

Naming a_x (b_y) the bit that Alice (Bob) sends once received the bit x (y), we can list all possible outcome of the CHSH game:

x	y	$x \wedge y$	$= a_x \oplus b_y$
0	0	0	$= a_0 \oplus b_0$
0	1	0	$= a_0 \oplus b_1$
1	0	0	$= a_1 \oplus b_1$
1	1	1	$= a_1 \oplus b_1$

We can see that, for a *deterministic classical strategy*, we have

$$\mathbf{Pr}\{\text{win}\} = \frac{1}{4} \sum_{a,b,x,y} V(x,y,a,b) p_{AB|XY}(a,b|x,y) \leq \frac{3}{4} \quad (7)$$

And that

$$\alpha := \mathbf{Pr}\{\text{win}\} - \mathbf{Pr}\{\text{lose}\} \implies \|4\alpha\| \leq 2 \quad (8)$$

Quantum Strategy [1]

Parameter λ can correspond to a shared quantum state $|\phi\rangle_{AB}$.

Alice's x -dependent local projective measurements: $\{\Pi_a^{(x)}\} : \sum_a \Pi_a^{(x)} = I$

Bob's y -dependent local projective measurements: $\{\Pi_b^{(y)}\} : \sum_b \Pi_b^{(y)} = I$

We then have

$$p_{AB|XY}(a, b|x, y) = \langle \phi|_{AB} \Pi_a^{(x)} \otimes \Pi_b^{(y)} |\phi\rangle_{AB} \quad (9)$$

and we can re-write Eq. 4 as

$$\mathbf{Pr}\{\text{win}\} = \frac{1}{4} \sum_{a,b,x,y} V(x, y, a, b) \langle \phi|_{AB} \Pi_a^{(x)} \otimes \Pi_b^{(y)} |\phi\rangle_{AB} \quad (10)$$

We define $A^{(x)} := \Pi_0^{(x)} - \Pi_1^{(x)}$ and $B^{(y)} := \Pi_0^{(y)} - \Pi_1^{(y)}$ and:

$$C_{AB} := A^{(0)} \otimes B^{(0)} + A^{(0)} \otimes B^{(1)} + A^{(1)} \otimes B^{(0)} - A^{(1)} \otimes B^{(1)} \quad (11)$$

Quantum Strategy - II [1], [2]

It is possible to show that

$$\alpha = \mathbf{Pr}\{\text{win}\} - \mathbf{Pr}\{\text{lose}\} = \frac{1}{4} \langle \phi |_{AB} C_{AB} | \phi \rangle_{AB} \quad (12)$$

and also that

$$\|C_{AB}\|_{\infty} \leq 2\sqrt{2} \quad (13)$$

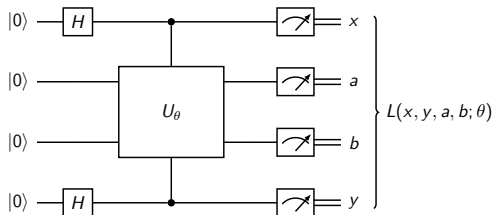
This implies that

$$\|4\alpha\|_{\infty} \leq 2\sqrt{2} \quad (14)$$

which is known as the *Tsirelson's bound*. This result translates to a larger maximum winning probability:

$$\mathbf{Pr}\{\text{win}\} \leq \frac{1}{2} + \frac{\sqrt{2}}{4} \approx 0.835 \quad (15)$$

General Idea



- ▶ The idea is to learn an optimal unitary such that some loss function is maximized.
- ▶ Relate the loss function to the possibility of winning the CHSH game.

An obvious choice is just the winning frequency on N simulations:

$$L(x, y, a, b; \theta) = \frac{1}{N} \sum_{i=1}^N V_\theta(x, y, a, b) \quad (16)$$

Parameterized Quantum Circuits [3]

$$|\psi_0\rangle \equiv \boxed{U_\theta} \equiv |\psi_\theta\rangle$$

We define a parameterized quantum circuit (PQC) as a parameterized unitary operation on n qubits, U_θ :

$$|\psi_\theta\rangle = U_\theta |\psi_0\rangle \quad (17)$$

where, usually, $|\psi_0\rangle = |0\rangle^{\otimes n}$. Sometimes compared to neural networks:

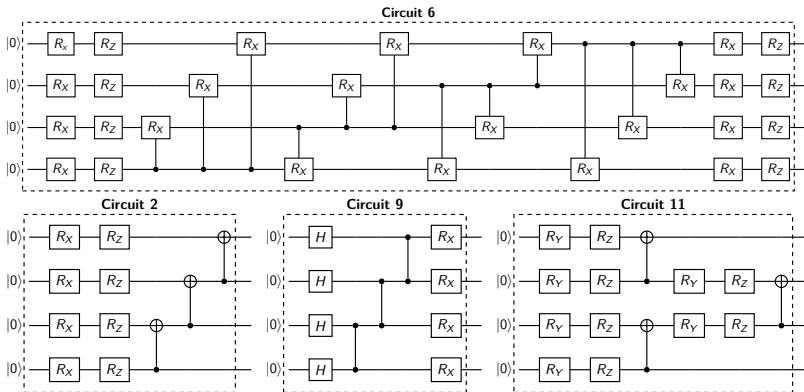
Similarities

- ▶ Universal approximator.
- ▶ Number of parameters scales polynomially in n .

Differences

- ▶ Unitary.
- ▶ Internal state not accessible.

Parameterized Quantum Circuits: Some Examples [4]



Expressibility and **Entangling Capability** studied on 19 (4-qubits) circuits wrt the increasing number of consecutive layers (from 1 to 4).

Circuit Descriptors: Entangling Capability [4], [5]

The entangling capability of a PQC is estimated through the Meyer-Wallach measure, here presented in the form given by Brennen. For a state $|\psi\rangle \in \mathbb{C}^{2^n}$ we define

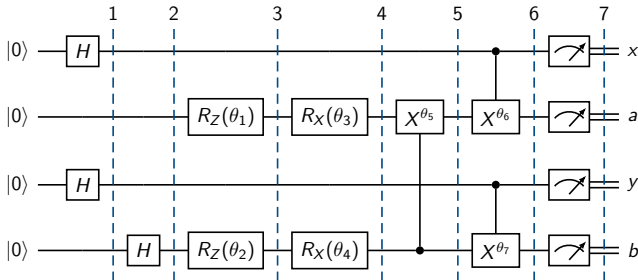
$$Q(|\psi\rangle) := \frac{1}{n} \sum_{k=1}^n 2(1 - \text{Tr}[\rho_k^2]) \quad (19)$$

where ρ_k is the one-qubit reduced density operator of the k -th qubit after tracing out the rest. Being $|\psi_\theta\rangle = U_\theta |\psi_0\rangle$ we define the entangling capability as

$$\text{Ent} := \frac{1}{|S|} \sum_{\theta \in S} Q(|\psi_\theta\rangle) \quad (20)$$

where $S = \{\theta_i\}_{i=0}^N$ is the set of sampled circuit parameters.

The PQC in detail



Parameters $\theta_1, \theta_2, \theta_3$ and θ_4 have been constrained inside $[0, 2\pi]$ while θ_5, θ_6 and θ_7 inside $[-2, 2]$.

Cost: There are 10 gates: 7 single-qubit gates and 3 two-qubit gates. The parametric gates are 4 rotations and 3 CNOTs.

Expressibility: $\text{Exp} \approx 0.03$.

Entangling capability: $\text{Ent} \approx 0.35$.

Bayesian Optimization [6]

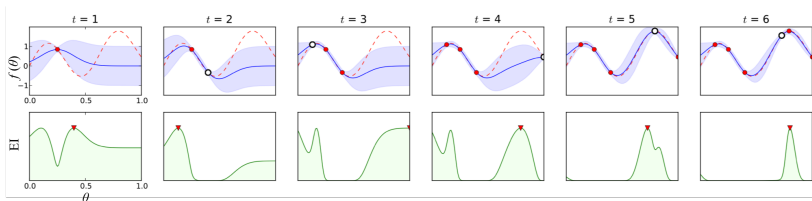
Aim: Find

$$\theta^* = \arg \max_{\theta \in \Theta} L(x, y, a, b; \theta) \quad (21)$$

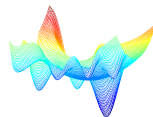
Idea: Attack the problem with Bayesian Optimization:

- ▶ Objective function is costly to evaluate.
- ▶ We don't have access to derivatives.
- ▶ The parameter space is "small" (usually $\dim(\Theta) < 20$).

Actors involved: a prior distribution, a posterior distribution and an *acquisition function*.



Simulation Setup

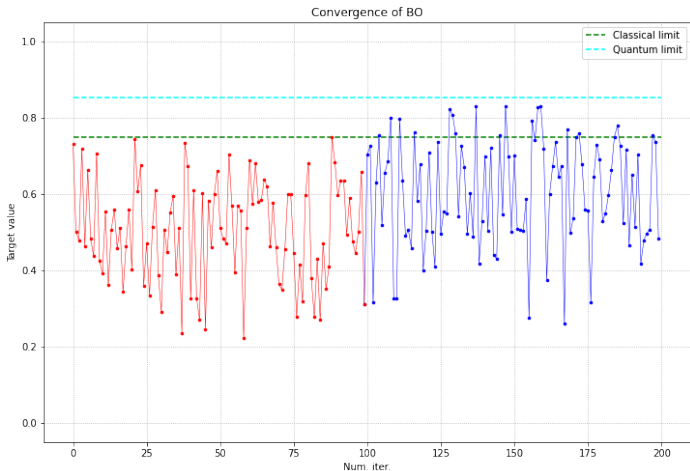


Pseudo-code

- 1: Init PQC
- 2: Init Simulator
- 3: **for** $i = 1, \dots, N$ **do**
- 4: Sample $\theta_i \in \Theta$ according to the acquisition function.
- 5: **for** $j = 1, \dots, M$ **do**
- 6: Run the simulation and collect data.
- 7: **end for**
- 8: Update prior and posterior based on data.
- 9: **end for**
- 10: **return** Best set of parameters θ^* .

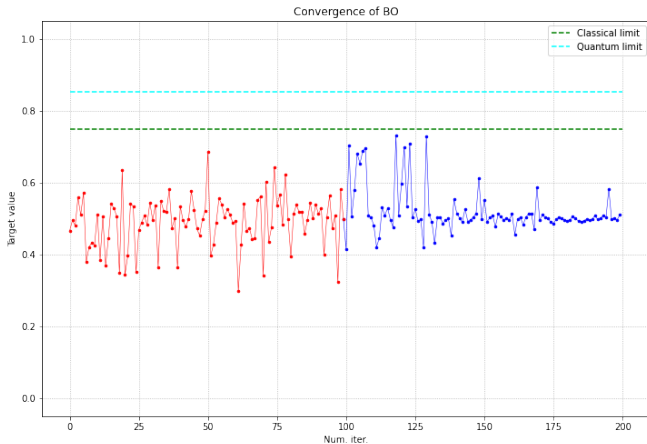
Results

Using $N = 200$ optimization steps and $M = 5000$ circuit simulations per step:



Results - II

Run the previous algorithm with a circuit with no entangling capability ($\text{Ent} = 0.$) and comparable expressibility ($\text{Exp} \approx 0.04$). This PCQ has 22 gates: 20 one-qubit gates and 2 two-qubit gates.



Conclusions

- ▶ Joint work of PQC and classical optimization → win the CHSH game
- ▶ Goal: show that entangle can emerge naturally from the setup
- ▶ More of a PoC than a real experiment
- ▶ Provide a simple setup for future and more complex ideas

Look at the code

https://github.com/w00zie/pqc_chsh



References I

- [1] M. M. Wilde, “Preface to the second edition,” *Quantum Information Theory*, pp. xi–xii, DOI: 10.1017/9781316809976.001. [Online]. Available: <http://dx.doi.org/10.1017/9781316809976.001>.
- [2] B. S. Cirel'Son, “Quantum generalizations of Bell's inequality,” *Letters in Mathematical Physics*, vol. 4, no. 2, pp. 93–100, Mar. 1980. DOI: 10.1007/BF00417500.
- [3] M. Benedetti, E. Lloyd, S. Sack, and M. Fiorentini, “Parameterized quantum circuits as machine learning models,” *Quantum Science and Technology*, vol. 4, no. 4, p. 043 001, Nov. 2019, ISSN: 2058-9565. DOI: 10.1088/2058-9565/ab4eb5. [Online]. Available: <http://dx.doi.org/10.1088/2058-9565/ab4eb5>.

References II

- [4] S. Sim, P. D. Johnson, and A. Aspuru-Guzik, “Expressibility and entangling capability of parameterized quantum circuits for hybrid quantum-classical algorithms,” *Advanced Quantum Technologies*, vol. 2, no. 12, p. 1900070, Oct. 2019, ISSN: 2511-9044. DOI: 10.1002/qute.201900070. [Online]. Available: <http://dx.doi.org/10.1002/qute.201900070>.
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- [6] P. I. Frazier, *A tutorial on bayesian optimization*, 2018. arXiv: 1807.02811 [stat.ML].

Maximum Quantum Winning Probability - 1

When the inputs x and y are 00, 01, 10 Alice and Bob win if they report back the same result. The probability is

$$\langle \phi |_{AB} \Pi_0^{(x)} \otimes \Pi_0^{(y)} | \phi \rangle_{AB} + \langle \phi |_{AB} \Pi_1^{(x)} \otimes \Pi_1^{(y)} | \phi \rangle_{AB}$$

The probability for it not to happen is

$$\langle \phi |_{AB} \Pi_0^{(x)} \otimes \Pi_1^{(y)} | \phi \rangle_{AB} + \langle \phi |_{AB} \Pi_1^{(x)} \otimes \Pi_0^{(y)} | \phi \rangle_{AB}$$

So in this case $\mathbf{Pr}\{\text{win}\} - \mathbf{Pr}\{\text{lose}\} = \langle \phi |_{AB} A^{(x)} \otimes B^{(y)} | \phi \rangle_{AB}$

When $x = 1$ and $y = 1$ we have that

$$\mathbf{Pr}\{\text{win}\} - \mathbf{Pr}\{\text{lose}\} = - \langle \phi |_{AB} A^{(1)} \otimes B^{(1)} | \phi \rangle_{AB}$$

So, averaging over all values of the input bits

$$\alpha := \mathbf{Pr}\{\text{win}\} - \mathbf{Pr}\{\text{lose}\} = \frac{1}{4} \langle \phi |_{AB} C_{AB} | \phi \rangle_{AB}$$

Maximum Quantum Winning Probability - 2

We have

$$C_{AB}^2 = 4I_{AB} - [A^{(0)}, A^{(1)}] \otimes [B^{(0)}, B^{(1)}]$$

and

$$\begin{aligned} \|C_{AB}^2\|_{\infty} &= \|4I_{AB} - [A^{(0)}, A^{(1)}] \otimes [B^{(0)}, B^{(1)}]\|_{\infty} \\ &\leq 4\|I_{AB}\|_{\infty} + \|[A^{(0)}, A^{(1)}] \otimes [B^{(0)}, B^{(1)}]\|_{\infty} \\ &= 4 + \|[A^{(0)}, A^{(1)}]\|_{\infty} \|[B^{(0)}, B^{(1)}]\|_{\infty} \\ &\leq 4 + 2 \cdot 2 = 8 \end{aligned}$$

So $\|C_{AB}\|_{\infty} \leq 2\sqrt{2}$ and

$$\|4\alpha\|_{\infty} = \|\langle \phi |_{AB} C_{AB} | \phi \rangle_{AB}\|_{\infty} \leq 2\sqrt{2}$$

Expressibility Calculation

Expressibility is

$$\text{Exp} := D_{\text{KL}}(P_{\text{PQC}}(F; \theta) \| P_{\text{Haar}}(F))$$

where, being $d = \dim(\mathcal{H})$

$$P_{\text{Haar}}(F) = (d - 1)(1 - F)^{d-2} \quad (22)$$

Expressibility

- 1: Init PQC
- 2: Init Simulator
- 3: **for** $i = 1, \dots, M$ **do**
- 4: Sample two sets of parameters θ_i, θ_j uniformly in Θ .
- 5: Calculate fidelity between the two output states.
- 6: **end for**
- 7: Calculate KL between fidelities and (22)

Expressibility Calculation - 2

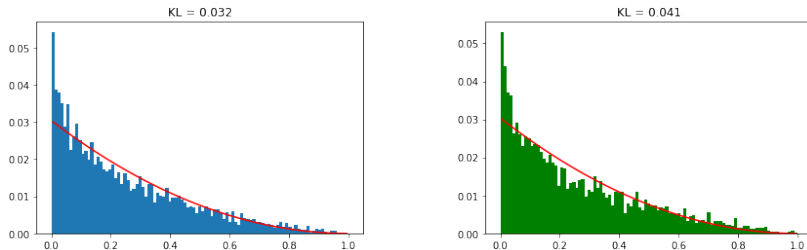


Figure 1: Entangling circuit (left) and non-entangling circuit (right)

Estimated with $M = 5000$ samples per circuit, compared to the theoretical density function (22) in red.