Parameterized Quantum Circuits and Bayesian Optimization for violating the Bell Inequality

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Introduction •0000000

Aim

Win a formulation of the CHSH game

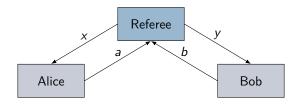
Methodology

Training a simple parameterized quantum circuit

Data

Outputs of quantum circuits simulations

Problem Formulation: CHSH Game [1]



• 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:

$$x \wedge y = a \oplus b \tag{1}$$

 $^{^1}$ Where \land denotes the logical \emph{AND} and \oplus the logical \emph{XOR} operations.

Strategy [1]

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> 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 $V: \{0,1\}^4 \to \{0,1\}$ as:

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

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

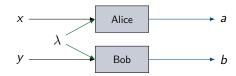
$$\Pr{\{\text{win}\}} = \sum_{a,b,x,y} V(x,y,a,b) p_{AB|XY}(a,b|x,y) p_{XY}(x,y) =$$

$$= \frac{1}{4} \sum_{a,b,x,y} V(x,y,a,b) p_{AB|XY}(a,b|x,y)$$
(3)

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

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

Classical Strategy [1]



We have conditional independence:

Also it is always possibile to write:

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

$$p_{AB|XY} = \int \int \int d\lambda \, dn \, dm \, f(a|\lambda, x, n) g(b|\lambda, y, m) p_{\Lambda}(\lambda) p_{N}(n) p_{M}(m)$$
$$= \int d\lambda \, f'(a|\lambda, x) g'(b|\lambda, y) p_{\Lambda}(\lambda)$$
(5)

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

Classical Strategy - II [1]

We can then write

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

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

Deterministic Classical Strategy [1]

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> 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	у	$x \wedge y$	$=a_x\oplus b_y$
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

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

And that

$$\alpha := \Pr{\min} - \Pr{\{\text{lose}\}} \implies \|4\alpha\| \le 2 \tag{8}$$

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

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

and we can re-write Eq. 4 as

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

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

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

Quantum Strategy - II [1], [2]

It is possible to show that

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

and also that

$$||C_{AB}||_{\infty} \le 2\sqrt{2} \tag{13}$$

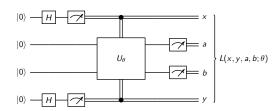
This implies that

$$||4\alpha||_{\infty} \le 2\sqrt{2} \tag{14}$$

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

$$\Pr{\min} \le \frac{1}{2} + \frac{\sqrt{2}}{4} \approx 0.835$$
 (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:

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

Parameterized Quantum Circuits [3]

$$|\psi_0\rangle$$
 U_{θ} $|\psi_{\theta}\rangle$

We define a parameterized quantum circuit (PQC) as a parameterized unitary operation on n qubits, U_{θ} :

$$|\psi_{\theta}\rangle = U_{\theta} |\psi_{0}\rangle \tag{17}$$

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

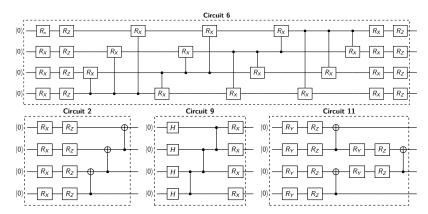
Similarities

- Universal approximator.
- Number of parameters scales polinomially in n.

Differences

- Unitary.
- Internal state not accessible.

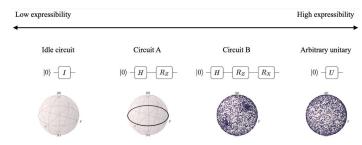
Parameterized Quantum Circuits: Some Examples [4]



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

Circuit Descriptors: Expressibility [4]

Idea: Quantify the ability of a PQC to generate states of an Hilbert space. The simplest case being $\mathcal{H} = \mathbb{C}^2$:



Estimated through the KL divergence between two distributions of fidelities: the first one coming from the PQC while the latter from the Haar ensemble:

$$\mathsf{Exp} := D_{\mathsf{KL}}(P_{\mathsf{PQC}}(F;\theta) \| P_{\mathsf{Haar}}(F)) \tag{18}$$

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 $|\psi\rangle\in\mathbb{C}^{2n}$ we define

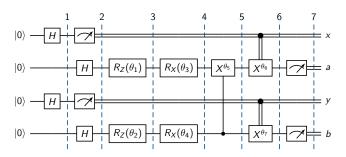
$$Q(|\psi\rangle) := \frac{1}{n} \sum_{k=1}^{n} 2(1 - \text{Tr}[\rho_k^2])$$
 (19)

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

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

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 11 gates: 8 single-qubit gates and 3 two-qubit gates.

The parametric gates are 4 rotations and 3 CNOTs.

Expressibility: Exp ≈ 0.05 .

Entangling capability: Ent ≈ 0.2 .

Bayesian Optimization [6]

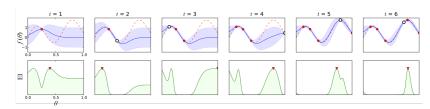
Aim: Find

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





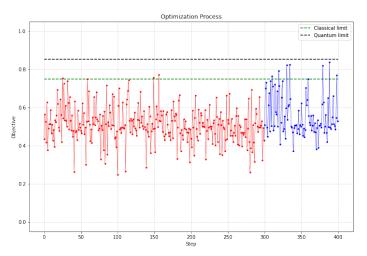


Pseudo-code

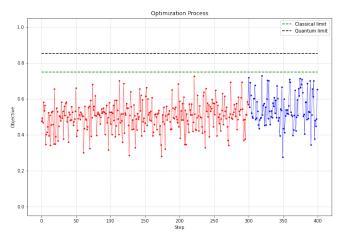
- 1: Init PQC
- 2: Init Simulator
- 3: **for** i = 1, ..., N **do**
- Sample $\theta_i \in \Theta$ according to the acquisition function. 4.
- for $i = 1, \dots, M$ do
- Run the simulation and collect data.
- end for
- Update prior and posterior based on data.
- 9: end for
- 10: **return** Best set of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

Using N=400 optimization steps and M=1000 circuit simulations per step:

Experiments & Results 0000000



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



Conclusions

- ightharpoonup Joint work of PQC and classical optimization \rightarrow 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

Experiments & Results 000000

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- [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.

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- [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.
- G. K. Brennen, "An observable measure of entanglement for pure [5] states of multi-qubit systems," Quantum Info. Comput., vol. 3, no. 6, pp. 619-626, Nov. 2003, ISSN: 1533-7146.
- P. I. Frazier, A tutorial on bayesian optimization, 2018. arXiv: [6] 1807.02811 [stat.ML].

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

$$\left\langle \phi\right|_{AB}\Pi_{0}^{(\mathrm{x})}\otimes\Pi_{0}^{(\mathrm{y})}\left|\phi\right\rangle_{AB}+\left\langle \phi\right|_{AB}\Pi_{1}^{(\mathrm{x})}\otimes\Pi_{1}^{(\mathrm{y})}\left|\phi\right\rangle_{AB}$$

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

When x = 1 and y = 1 we have that

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

So, averaging over all values of the input bits

$$\alpha := \mathbf{Pr}\{\mathsf{win}\} - \mathbf{Pr}\{\mathsf{lose}\} = \frac{1}{^{\mathit{A}}} \left\langle \phi |_{\mathit{AB}} \, \mathit{C}_{\mathit{AB}} \, | \phi \right\rangle_{\mathit{AB}}$$

We have

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

and

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

$$\leq 4||I_{AB}||_{\infty} + ||[A^{(0)}, A^{(1)}] \otimes [B^{(0)}, B^{(1)}]||_{\infty}$$

$$= 4 + ||[A^{(0)}, A^{(1)}]||_{\infty} ||[B^{(0)}, B^{(1)}]||_{\infty}$$

$$\leq 4 + 2 \cdot 2 = 8$$

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

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

Expressibility Calculation

Expressibility is

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

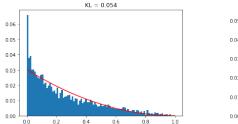
where, being $d = dim(\mathcal{H})$

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

Expressibility

- 1: Init PQC
- 2: Init Simulator
- 3: **for** i = 1, ..., 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



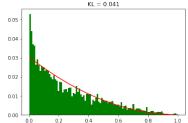


Figure 1: Entangling circuit (left) and non-entangling circuit (right)

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