

Computability theory and Bell non-locality

Antonio Acín¹, Ariel Bendersky², Gonzalo de La Torre¹,
Santiago Figueira² y *Gabriel Senno*¹

¹ICFO - The Institute of Photonic Sciences.

²Computer Science Department, FCEyN, University of Buenos Aires.



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Abstract

Two results relating computability theory and Bell non-locality:

- Inputs: *The computability loophole.*

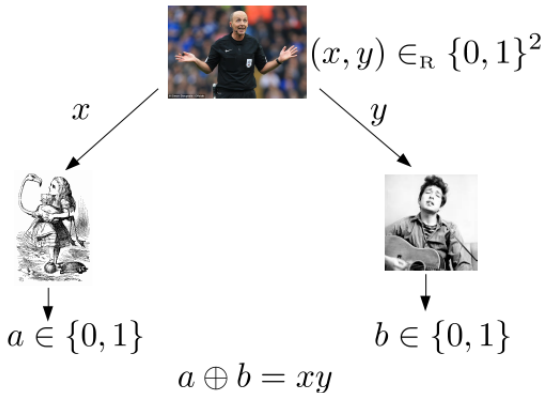
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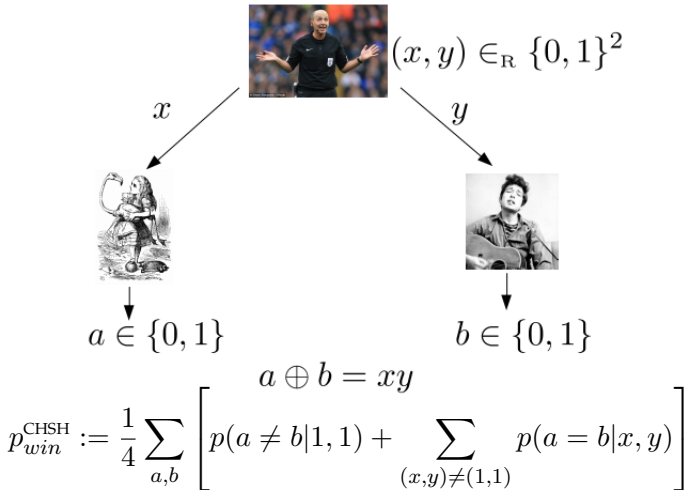
- Inputs: *The computability loophole.*
- Outputs: *Computability + non-locality \implies signaling.*

The computability loophole

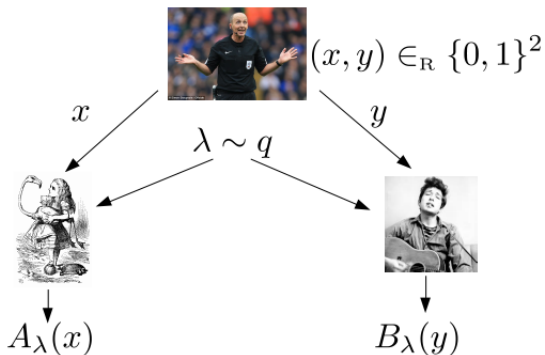
The CHSH game



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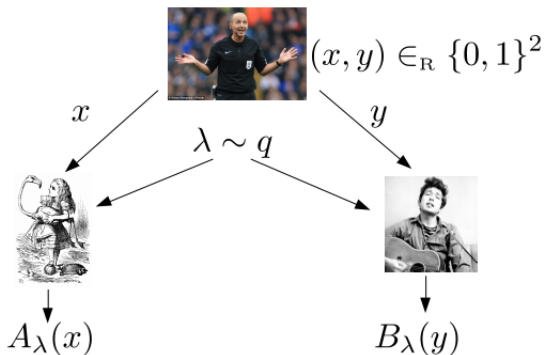


Local strategies



$$p(a, b|x, y) = \sum_{\lambda} q(\lambda) \delta_{a=A_{\lambda}(x)} \delta_{b=B_{\lambda}(y)}.$$

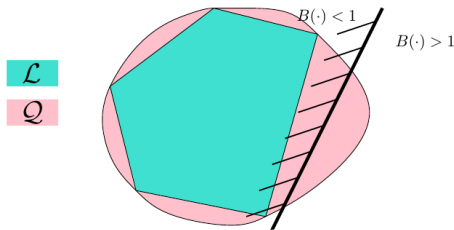
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$$p_{win}^{\text{CHSH}} \leq \frac{3}{4}, \text{ for every local strategy.}$$

Bell inequalities



- The CHSH inequality

$$p(= |P_1, P_1) + p(= |P_1, P_2) + p(= |P_2, P_1) + p(\neq |P_2, P_2) \leq 3$$

is an example of a **Bell inequality**.

- In general,

$$\sum_{a,b,x,y} B_{a,b,x,y} p(a,b|x,y) \leq B_l.$$

Quantum strategies

- Quantum strategy:

$$p(a, b|x, y) = \langle \psi | \Pi_a^x \Pi_b^y | \psi \rangle$$

with $|\psi\rangle \in \mathcal{H}$, $\sum_a \Pi_a^x = \sum_b \Pi_b^y = \mathbb{I}_{\mathcal{H}}$ and $[\Pi_a^x, \Pi_b^y] = 0$.

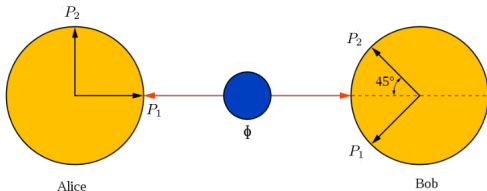
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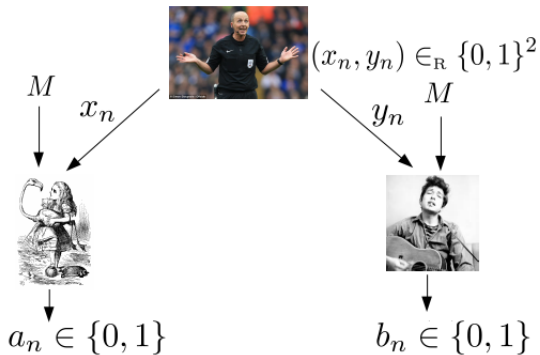
- For the CHSH game, preparing $|\psi^-\rangle =: \frac{1}{\sqrt{2}}(|01\rangle - |10\rangle)$ and measuring in the following spin directions:



it is easy to see that, $p_{win}^{CHSH} = \cos^2(\pi/8) \approx 0,85$.

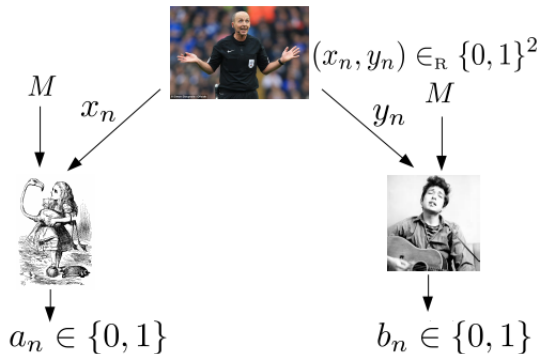
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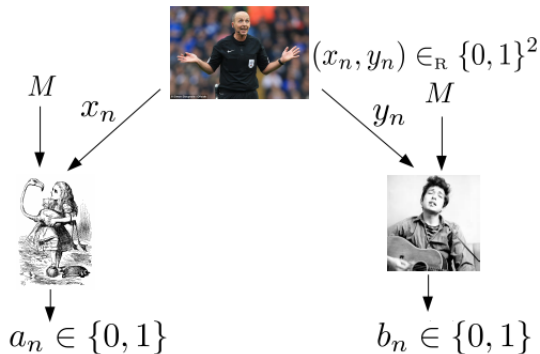


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$$p(a_n, b_n | x_n, y_n) = \sum_{\lambda} q(\lambda) \delta_{a_n = A_{\lambda}(x_n, M)} \delta_{b_n = B_{\lambda}(y_n, M)}.$$

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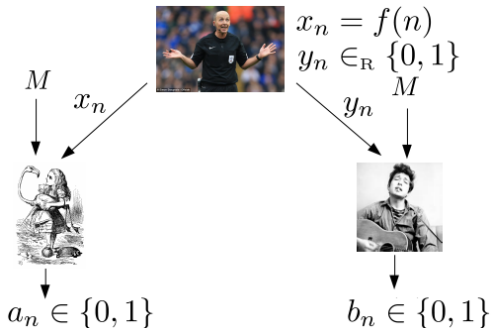
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- $p_{win}^{CHSH} \leq 3/4$ [Barrett et al., Phys. Rev. A 66, 042111, 2002].

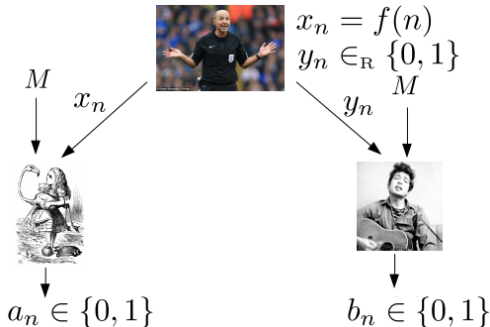
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Theorem ([Bendersky, Senno, de la Torre, Figueira and Acín. PRL 116, 230402, 2016])

If the referee in the CHSH game with memory chooses (at least) one of the players' questions using a computable function $f : \mathbb{N} \rightarrow \{0, 1\}$, there is a perfect local strategy (independent of f).

Computable functions

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- But, the class of all computable functions is *not* computably enumerable.

Predicting computable functions

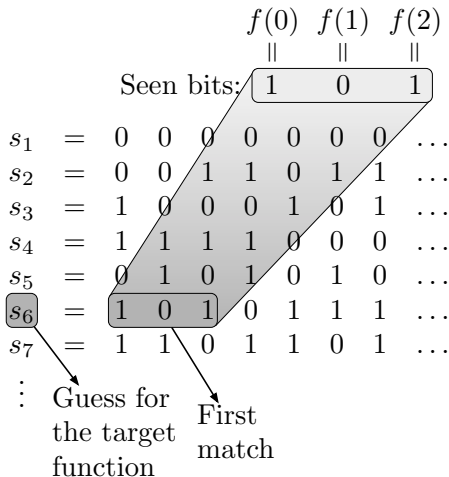
Functions in computably enumerable classes can be predicted in the following sense:

For every computably enumerable class \mathcal{C} of computable functions there is a program \mathcal{P} (called a *predictor for \mathcal{C}*) such that for every $f \in \mathcal{C}$,

$$(\exists n_0)(\forall n \geq n_0) f(n) = \mathcal{P}([f(0), \dots, f(n-1)])$$

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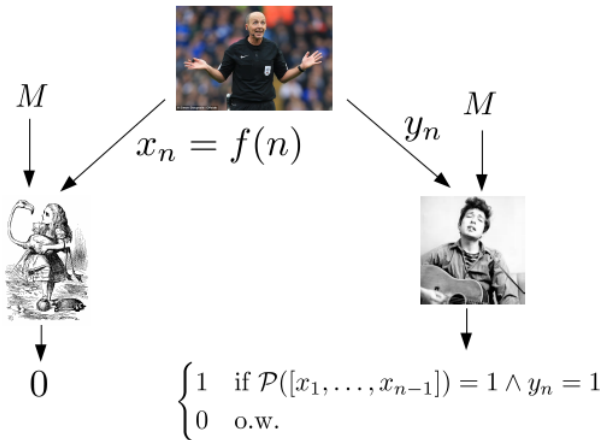
The predictor works as follows,



Perfect local strategy

Let $T : \mathbb{N} \rightarrow \mathbb{N}$ be some computable function, $f \in \mathcal{C}_T$ and \mathcal{P} a predictor for \mathcal{C}_T .

$$M = [(x_1, y_1), \dots, (x_{n-1}, y_{n-1})]$$



Technical details of the model

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- Nevertheless, the number M of prediction errors he make until that time is small.
 - If $f(n)$ has a k -bits program, $M \leq O(\log(k))$.
- Also, the prediction algorithm is (almost) as efficient as f .
 - If f is computable in $O(T)$ time, then \mathcal{P} runs in $O(T \cdot \log(T))$.

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 - Rather circular to assume a non-local theory when testing non-locality.
 - Experimentally unverifiable.
- Nevertheless, the result I will talk about next implies that, under reasonable assumptions, the outputs from QRNGs are, in fact, uncomputable.

Computable non-locality allows
for signaling

Deterministic boxes in a CHSH scenario

Round n

$$x \in \{0, 1\}$$



$$a = A(x, n) \in \{0, 1\}$$

$$y \in \{0, 1\}$$



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$$p(a, b|x, y) := \lim_{n \rightarrow \infty} \frac{|\{i \leq n \mid x_i = x, y_i = y, a_i = a, b_i = b\}|}{|\{i \leq n \mid x_i = x, y_i = y\}|}.$$

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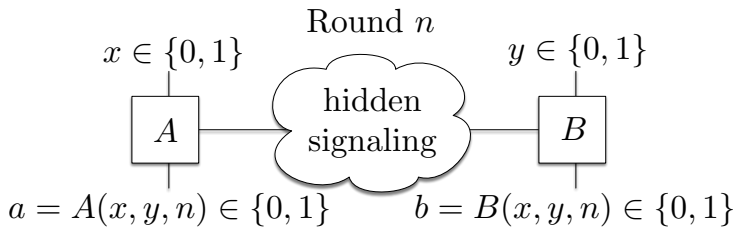


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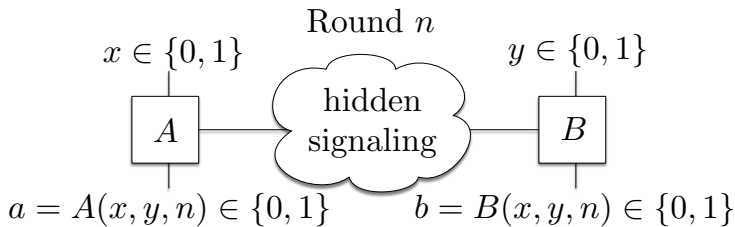
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We will say that A and B are *non-local* if, when the inputs are chosen uniformly at random, p violates a Bell inequality with probability 1.

Determinism \wedge non-locality \implies violation of Parameter Independence



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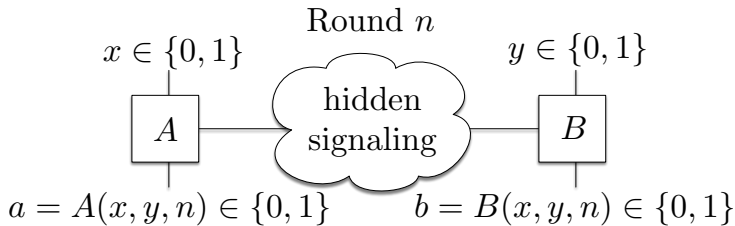


Lemma

If A and B are non-local,

$$\exists^\infty n [\exists x A(x, 0, n) \neq A(x, 1, n) \vee \exists y B(0, y, n) \neq B(1, y, n)].$$

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Observation

Violation of Parameter Independence doesn't imply signaling
(e.g.: Bohmian mechanics, Toner & Bacon, etc).

Using that hidden signaling for communicating

W.l.o.g., let's assume that

$$\exists^\infty n \exists y B(0, y, n) \neq B(1, y, n). \quad (1)$$

Observation

If Alice and Bob had access to B , i.e. if they knew how to **compute** it, they could easily communicate: they just wait for the n s that verify (1) and, with the right choice of y_n , Bob can tell x_n . Thus, we assume B is hidden (it's Nature's secret). Can it be kept that way?

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Main Result

We give a protocol which, if B is a **computable function**, allows Alice to send a message to Bob with the sole knowledge of a bound on the computational complexity of B .

Learnability in the limit

A class computable functions \mathcal{C} is *learnable in the limit* if there exists a program \mathcal{P} (called a *learner for \mathcal{C}*) such that for every $f : \mathbb{N} \rightarrow \mathbb{N} \in \mathcal{C}$, there exists m such that for every $m \geq n$, on input (some coding of) $\{(0, f(0)), \dots, (m, f(n))\}$ \mathcal{P} outputs (the code of) a program that computes f .

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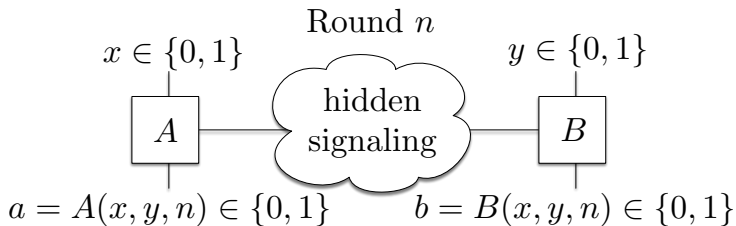
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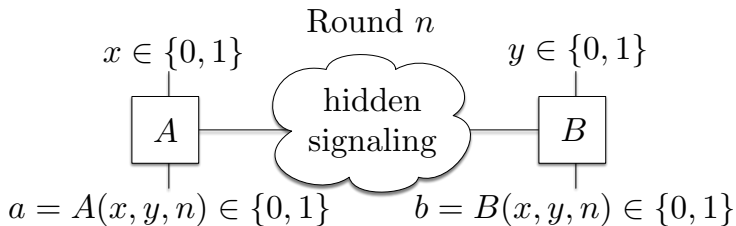
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- Every computably enumerable class of computable functions is learnable in the limit.
- The class of all computable functions is not learnable in the limit.

Learning in the limit allows communication



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Restrictions

- In order for Bob to learn a program to compute the function B , he needs to know Alice's inputs x , at least until B has been learned.
- For every n , Bob will only see the value of B for just one pair of inputs (x_n, y_n) .
- Bob will not be able to tell when he has effectively learned B .

The protocol $\mathcal{P}(t, m, S)$

Inputs:

- 1 a computable non-decreasing function t
- 2 the size of Alice's message m
- 3 a sequence $S \in \{(0, 0), (0, 1), (1, 0), (1, 1), 1, \dots, m\}^\infty$

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On each round n :

- 1 if $S(n) = (x, y)$, Alice inputs x and Bob inputs y . Then, Bob uses a learner for the class of functions computable in time $O(t)$ on input $((x_{i_1}, y_{i_1}, B(x_{i_1}, y_{i_1}, i_1)) \dots (x, y, B(x, y, n)))$ (with i_k being the past learning rounds) to update his guess \tilde{B} of a program that computes B (**Learning round**).
- 2 if $S(n) = i \in \{1, \dots, m\}$, Alice inputs the i th bit of her message, a_i and Bob y s.t. $\tilde{B}(0, y, n) \neq \tilde{B}(1, y, n)$ and makes the output of his box his new guess for a_i . If there is no such y , he inputs 0 (**Signaling round**).

Soundness of the protocol

For $\mathcal{P}(t, m, S)$ to be sound, it suffices that the following properties hold:

- 1 There exists a number of round n such that for all $m \geq n$, and $x, y \in \{0, 1\}$, we have $\tilde{B}(x, y, m) = B(x, y, m)$, i.e. the learning process converges to B .
- 2 For every bit i of Alice's message and for infinitely many n , $S(n) = i \in \mathbb{N}$ and $\exists y \in \{0, 1\}. B(0, y, n) \neq B(1, y, n)$.

Alternating randomly

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Can we de-randomize?

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Theorem ([Bendersky, Senno, de la Torre, Figueira, Acín. Phys. Rev. Lett. 118, 130401, 2017])

If S is T -random, properties 1 and 2 hold.

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- Algorithmic randomness cannot be *computably* amplified [Miller, Adv. Math. 226(1), 373-384 (2011)].

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2 Computability of the set of quantum correlations.

- Very recent breakthrough results by Slofstra [arXiv:1606.03140, arXiv:1703.08618].