Nonlocality can be quantified by the violation of a Bell inequality. Since this violation may be amplified by local operations an alternative measure has been proposed – distillable nonlocality. The alternative measure is difficult to calculate exactly due to the double exponential growth of the parameter space. In this article we give a way to bound the distillable nonlocality of a resource by the solutions to a related optimization problem. Our upper bounds are exponentially easier to compute than the exact value and are shown to be meaningful in general and tight in some cases.

I. INTRODUCTION

When two separated parts of a quantum state are measured in different bases, then the outcomes can be correlated in a way that cannot be explained by information shared before the separation [1]. This property has been termed quantum nonlocality. Nonlocal correlations have since proved to be a useful resource leading to new applications such as, for example, quantum key distribution [2, 3]. Stronger correlations that are still in accordance with the nonsignaling postulate of relativity [6] can be defined and are formalized and studied in so-called generalized nonsignaling theories [4, 5] in which quantum correlations are a special case. Assuming nonlocality that is super-quantum to some extent has interesting consequences for nonlocal computation [9] and for communication complexity [10, 11]. Maximal nonlocality allows to compute every distributed Boolean function with just one communicated bit [12].

Local correlations obey certain linear constraints, so-called Bell inequalities [1]. The extent by which a Bell inequality is violated by a correlation can be taken as a measure for nonlocality (see [13–15] for other nonlocality measures and related results). The question arose whether this measure is also meaningful for quantifying the nonlocality of a resource consisting of several copies of a correlation. In a context where nonlocality is more useful the stronger it is this measure is problematic if stronger nonlocality can be obtained from a number of weakly nonlocal correlations. In other words, we want to know if two parties having access to correlations violating a Bell inequality by some small extent can execute local operations to obtain a higher violation. The two parties can carry out arbitrary local operations, but cannot communicate. This process is called nonlocality distillation.

It has recently been shown that the Clauser-Horner-Shimony-Holt inequality [16] (CHSH) as a measure for nonlocality in minimal dimensions is indeed problematic because it is distillable in general [17]. A protocol by Brunner and Skrzypczyk [11] manages to distill an arbitrarily weak nonlocal correlation, that is still super-quantum, to the extent where communication complexity collapses. As proposed in [17] a more meaningful measure, which is by definition undistillable, is the maximal CHSH violation achievable from many realizations of a given nonlocal resource by any distillation protocol. Brunner et al. [18] recently compared this alternative measure, termed distillable nonlocality, to the Elitzur-Popescu-Rohrlich (EPR2) decomposition approach [14, 15]. The authors discovered examples of bound nonlocality and activation – the box-world analogues to bound entanglement and its activation.

Haye and Rashid [19] thoroughly analyzed different distillation protocols, proved optimality in restricted classes and discovered a strong dependence between the optimal protocol and the parameters of the resource that it distills. Today our knowledge about general distillation protocols is still very limited. Although tools and techniques to analyze fixed protocols exist (for example by discrete maps [11, 20]), difficulties arise and the lack of a neat framework becomes obvious when attempting to calculate the distillable nonlocality of a fixed resource. Besides what results from some trivial symmetries not much is known to simplify the task. It usually requires an exhaustive search over all possible distillation protocols. The fact that the search space grows doubly exponential in the number of involved correlation copies represents a serious problem for the usefulness and understanding of this alternative nonlocality measure. Therefore, as recently suggested in [13], meaningful bounds for distillable nonlocality that can be derived more efficiently are a reasonable compromise.

In this article we propose such bounds and analyze them. They rely on a recursive optimization problem that has a strong connection to the process of optimizing distillation protocols for isotropic nonlocality. Solved with a dynamic-programming approach our bounds yield an exponential gain in the run-time compared to a brute-force search. The efficiency advantage enables one to analyze instances with the number of available correlations limited by 9. To demonstrate this we calculated the distillable nonlocality bounds for a set of fixed isotropic systems and found that, in this context, our bounds rule out distillation and are therefore tight. Furthermore, we present a general idea how to extend isotropic bounds to the non-isotropic case. Here, our calculated bounds show that the distinguishable nonlocality approaches the CHSH nonlocality of a single copy when the resource is chosen closer to an isotropic line.
II. PRELIMINARIES

Following a general approach to nonlocality [3], we consider correlations in the joint behavior of the two ends of a bipartite input-output system, characterized by joint probability distributions \( P_{AB}^{xy} \) on random variables \( A, B \) for each \( x, y \). Let \( x \) and \( a \) be the input and output on the left-hand side of the system, and \( y \) and \( b \) the corresponding values on the right-hand side. On inputs \( x \) and \( y \) the system returns outputs \( a \) and \( b \) with probability \( P_{AB}^{xy}(a, b) \).

\[
x \rightarrow P_{AB}^{xy} \quad \leftarrow y
\]
\[
a \quad b
\]

In a distillation setting two parties, Alice and Bob, sharing a number of systems can independently choose inputs – possibly derived from local outputs of other systems – and collect outputs on their ends of the systems. In other words, they can apply any classical circuitry to their local parts of the systems. Such a local input-output strategy is called a wiring [3, 21]. A party receives its output from a system immediately after giving its input, independently of whether the other has given its input already. This prevents the parties from signaling by delaying their inputs. On the other hand it allows for wirings that have a different temporal order on the two sides. For example: While Alice’s input into system 2 depends on the output of system 1, Bob’s input into system 1 depends on the output of system 2. If the temporal order on both sides is equal we call the wiring ordered and disordered otherwise. In this paper we consider both kinds, summarized under general wirings as defined in [22]. Since we are only interested in optimality ignoring non-deterministic strategies is sufficient due to convexity. Thus, we are allowed to ignore randomness shared between the two players to simplify the proofs.

Here, we assume nonlocality in its simplest form, namely, in the binary setting where each input and each output has two possible values, i.e., \( a, b \in \{0, 1\} \) and \( x, y \in \{0, 1\} \). We refer to [3] for a detailed description of the convex polytope of binary nonsignaling systems (\( \mathcal{NS} \)). In the case where both inputs and both outputs are binary, the only Bell inequality (up to symmetries) is the CHSH inequality. Furthermore, the set of eight CHSH inequalities is complete for binary systems in the sense that if none of them is violated, then the system is local. A system is called isotropic if it is a probabilistic mixture of opposite nonlocal vertices of \( \mathcal{NS} \), which are known as PR-boxes [3]. See [23] for a formal definition of isotropic systems.

Let \( P \) abbreviate any binary system. Suppose that Alice and Bob share \( n \) copies of \( P \), which we index as \( P_1, ..., P_n \). The outputs of the \( i \)-th system \( P_i \) are denoted \((a_i, b_i) \in \{0, 1\}^2 \). So, the binary strings \( a = (a_1, a_2, ..., a_n) \in \{0, 1\}^n \) and \( b = (b_1, b_2, ..., b_n) \in \{0, 1\}^n \) are the outputs of the \( n \) systems to Alice and to Bob. Also, from now on, let \( A = (A_1, ..., A_n) \) and \( B = (B_1, ..., B_n) \) denote random variables for the strings of the \( n \) collected bits on both sides. Any wiring of \( n \) shared copies of a system then fully determines a joint distribution \( W_{AB} \) on the space of all sequences \((a, b) \in \{0, 1\}^{n \times n} \).

Alice and Bob can condition their wirings on the inputs \( x \) and \( y \). We denote a system of wiring distributions on \( n \) copies of the binary system \( P \) by \( P^n \). We will use the matrix representation

\[
P^n = \begin{bmatrix}
W_{AB}^{00} & W_{AB}^{01} \\
W_{AB}^{10} & W_{AB}^{11}
\end{bmatrix}
\]

where \( W_{AB}^{xy} \), for any \( x, y \in \{0, 1\} \), is a \( 2^n \times 2^n \) matrix with the probability \( W_{AB}^{xy}(a, b) \) at position \((a, b) \). As the final step in a distillation attempt the two parties map their inputs \( x \) and \( y \) and the collected strings \( a \) and \( b \) to one local bit each. For all \( x, y \) let \( f_x, g_y : \{0, 1\}^n \rightarrow \{0, 1\} \) stand for these local Boolean functions. It will often be convenient to write them as truth tables \( f_x, g_y \in \{0, 1\}^{2^n} \). Now we have all the ingredients to define a distillation protocol on binary systems.

Definition 1. A deterministic nonlocality distillation protocol on \( n \) copies of a system \( P \), denoted \( DP = (P^n, f_0, f_1, g_0, g_1) \), consists of a system of wiring distributions \( P^n \) and local decision functions \( f_0, f_1, g_0, g_1 \) for both inputs and both sides.

We refer to [10] for an exact formulation of the CHSH nonlocality and to [17] for the definition of the CHSH nonlocality of a single binary system \( NL(P) \) as required here. Note that \( NL(P) > 2 \) indicates that \( P \) can be used to violate a CHSH inequality and is therefore called nonlocal. To measure the violation of a CHSH inequality by a distillation protocol we use a slightly different representation. Observe that for each input pair \( x, y \) the correlation function of the binary system simulated by a distillation protocol can be described by the following inner product:

\[
E[f_x(a) = g_y(b)] - E[f_x(a) \neq g_y(b)] \quad = \sum_{ab} W_{AB}^{xy}(a, b)(1 - 2f_x(a))(1 - 2g_y(b)) \quad = (1 - 2f_x)^T W_{AB}^{xy}(1 - 2g_y).
\]

The two expectation values are calculated over all pairs \((a, b) \sim W_{AB}^{xy} \). The nonlocality of a protocol is the maximal violation of a CHSH inequality by the simulated system.

Definition 2. The CHSH value of a distillation protocol \( DP = (P^n, f_0, f_1, g_0, g_1) \) is measured by

\[
NL(DP) = \max_{xy} |E[f_x a] + E[f_x g] + E[f_bar a] - E[f_bar g]|
\]

where we use \( \bar{x} \) and \( \bar{y} \) to indicate bit flips, that is, \( \bar{0} = 1 \) and \( \bar{1} = 0 \). Note that \( NL(DP) > 2 \) indicates
that the protocol simulates a nonlocal system, whereas $NL(DP) > NL(P)$ indicates a successful distillation attempt.

We will henceforth assume that the maximum of this expression is reached at $xy = 00$. Our reasoning can easily be extended to all four cases.

Motivated by the discovery of successful distillation protocols for CHSH nonlocality another measure has been proposed that expresses the nonlocality of the optimal protocol on a given resource. Brunner et al. [18] have recently formally defined and analyzed distillable nonlocality in comparison with the EPR2 decomposition [14, 13]. Here, we use a slightly different notation that expresses the same idea.

**Definition 3.** The distillable nonlocality of $n$ copies of a system $P$ is defined as

$$D(n, P) = \max_{DP} NL(DP),$$

where we maximize over all deterministic nonlocality distillation protocols on $n$ copies of $P$.

For determining distillable nonlocality precisely by an exhaustive search over the space of all possible sequences $(P^n, f_0, f_1, g_0, g_1)$ one requires to test roughly

$$\left(\prod_{i=0}^{n} 2^{2^{i+1}}\right)^2 = 2^{\sum_{i=0}^{n} 2^{i+2}} \in 2^{O(n)}$$

instances, calculated as the product of all involved Boolean functions. Because this is infeasible with normal hardware already for $n > 2$, giving an asymptotic bound to $D(n, P)$ instead which can be computed more efficiently, is a reasonable compromise.

**III. RESULTS**

First, we show an upper bound on $D(n, P)$ if $P$ is an isotropic system (Corollary 1) and move on to a general upper bound afterward (Corollary 2). We start by manipulating (1) to reveal the core of the task. One can group the Boolean functions $f_0, f_1, g_0, g_1$, over which we optimize in (1), by their output distribution. For a given number $n$, a class $C_k$ of Boolean functions is defined as

$$C_k = \{ f : \{0, 1\}^n \rightarrow \{0, 1\} : |f^{-1}(1)| = k \}.$$  

We can now rewrite the distillable nonlocality as

$$D(n, P) = \max_{0 \leq k^x, l^y \leq 2^n} \max_{DP} NL(DP),$$

where for each $k^x, l^y$ we maximize over all distillation protocols restricted by $f_x \in C_{k^x}$ and $g_y \in C_{l^y}$ for all $x, y$. Now concentrate on $\max_{DP} NL(DP)$ with fixed $k^x$ and $l^y$ for all $x, y$. According to Definition 2 this subproblem boils down to finding the system of wiring distributions $P^n$ and functions $f_0 \in C_{k^{x_0}}, f_1 \in C_{k^{x_1}}, g_0 \in C_{l^{y_0}}, g_1 \in C_{l^{y_1}}$, such that the term

$$\langle f_0, g_0 \rangle + \langle f_1, g_0 \rangle + \langle f_0, g_1 \rangle - \langle f_1, g_1 \rangle$$

is minimal/maximal. Since for any $x, y$ we have

$$\langle f_x, g_y \rangle = (1 - 2f_x)W_{AB}^{xy}(1 - 2g_y)$$

$$= 1 - k^x/2^{n-1} - l^y/2^{n-1} + 4f_x W_{AB}^{xy}g_y,$$

bounds for the correlation functions $\langle f_x, g_y \rangle$ can be derived from optimizing $f^x W_{AB}g y$, with $W_{AB}$ a wiring distribution and $f \in C_{k^x}, g \in C_{l^y}$, independently for all $x, y$. We will derive bounds to $f^x W_{AB}g y$ by slightly relaxing the constraint that $W_{AB}$ needs to be a distribution obtained from a wiring. This relaxation has the following background:

Suppose fixed functions $f \in C_{k^x}, g \in C_{l^y}$ and a fixed wiring distribution $W_{AB}$. Let $(a_1, b_1)$ be the outputs of the first system Alice accesses in this wiring. The $k^x$ preimages of 1 under $f$ are split into two parts with reference to the output bit $a_1$ and the $l^y$ preimages of 1 under $g$ are split into two parts according to $b_1$, i.e., we have

$$k^x = \{|f^{-1}(1) : a_1 = 0\}| + \{|f^{-1}(1) : a_1 = 1\}|$$

and

$$l^y = \{|g^{-1}(1) : b_1 = 0\}| + \{|g^{-1}(1) : b_1 = 1\}|.$$  

The wiring $W_{AB}$ provides a probability distribution on $(a_1, b_1)$, possibly conditioned on outputs of some other shared systems. Roughly speaking we will find the optimal splitting of $f, g$ and the optimal distribution for the output pair $(a_1, b_1)$ by considering the optimal splittings and optimal distributions in the four subproblems, where the pair $(a_1, b_1)$ is assumed to have fixed values. In this way we obtain a recursive $n$-level hierarchy of optimization problems. The following definition makes this idea precise.

A problem instance is given by the tuple $(P, n, k, l)$, with a binary system $P$, integers $n \geq 1$ and $0 \leq k, l \leq 2^n$. The recursive optimization to solve is:

**Definition 4.** Given any instance $(P, n, k, l)$ fix the parameter $p = P^0_{AB}(0, 0)$ and find the values

$$\delta^+_n(k, l) = \left\{ \begin{array}{ll} \max_{ij} p[\delta_{n-1}^+(i, j) + \delta_{n-1}^+(k-i, l-j)] \\
 & + (1/2-p)[\delta_{n-1}^+(i, l-j) + \delta_{n-1}^+(k-i, j)], \\
 & \text{if } n > 0, \\
 & kl, \quad \text{otherwise,} \end{array} \right.$$  

and

$$\delta^-_n(k, l) = \left\{ \begin{array}{ll} \min_{ij} p[\delta_{n-1}^-(i, j) + \delta_{n-1}^-(k-i, l-j)] \\
 & + (1/2-p)[\delta_{n-1}^-(i, l-j) + \delta_{n-1}^-(k-i, j)], \\
 & \text{if } n > 0, \\
 & kl, \quad \text{otherwise,} \end{array} \right.$$  

where we optimize over the ranges $k - \min(k, 2^{n-1}) \leq i \leq \min(k, 2^{n-1})$ and $l - \min(l, 2^{n-1}) \leq j \leq \min(l, 2^{n-1})$. 


Solutions $\delta^+_n(k^x,l^y)$ and $\delta^-_n(k^x,l^y)$ to the instance $(P,n,k^x,l^y)$ capture the possibilities we have when constructing a wiring distribution $W_{AB}$ and functions $f \in C_{k^x}$, $g \in C_{l^y}$, such that the product $f^gW_{AB}$ is optimal. Since in the above optimization the subproblems on the same level are handled independently we get better solutions than with distributions from real wirings, where this independence is not given.

**Lemma 1.** Let $P$ be any isotropic system and suppose integers $0 \leq k, l \leq 2^n$. For any wiring distribution $W_{AB}$ on $n$ copies of $P$ and functions $f \in C_{k^x}, g \in C_{l^y}$ we have

$$\delta_n^-(k,l) \leq f^gW_{AB} \leq \delta_n^+(k,l),$$

where $\delta_n^-(k,l)$ and $\delta_n^+(k,l)$ are the solutions to $(P,n,k,l)$.

**Proof.** The idea for proving the upper bound is to decompose $W_{AB}$ into four subwirings and $f,g$ into two functions each, on which we then inductively prove the statement. The decomposition is done with reference to the first system Alice accesses, called $P_1$, its outputs denoted $(a_1,b_1)$. We write $a_1$ for the string $(a_2,\ldots,a_n)$ and $b_1$ for $(b_2,\ldots,b_n)$ and accordingly we use the random variables $A_1 = (A_2,\ldots,A_n)$ and $B_1 = (B_2,\ldots,B_n)$.

We fix the probability $P = P_{AB}^0(0,0)$. Let the Boolean function $h : \{0,1\}^{n-1} \rightarrow \{0,1\}$ determine a bit depending on Bobs local outputs $b_1 = (b_2,\ldots,b_n)$ of the $n-1$ systems $P_2,\ldots,P_n$ shared with Alice, such that

$$W_{A_1B_1|B_1}(0,h(b_1)) = p$$

for all $b_1 \in \{0,1\}^{n-1}$. Note that if the wiring defining $W_{AB}$ is ordered then Alice’s and Bobs inputs into the first system are constant and thus $h$ is also constant. Otherwise, only Alice’s input is guaranteed to be constant. In this case $h$ is used to react to possible changes caused by Bobs input into the first system. $h$ always exists since isotropic systems have symmetric output distributions, and therefore for any $b_1$ the first system $P_1$ outputs either $(0,0)$ or $(0,1)$ with probability $p$.

According to the result of $h$ we split $W_{AB}$ into four parts by grouping all probabilities $W_{AB}(a,b)$ by four possible outcome pairs of the first system. In the same manner we split the functions $f,g$ into the subfunctions $f'(a_1) = f(0,a_1)$ and $f''(a_1) = f(1,a_1)$ for any $a_1$ and $g'(b_1) = g(h(b_1),b_1)$ and $g''(b_1) = g(h(b_1),b_1)$ for any $b_1$.

Applied to $f^gW_{AB}$ we can identify four parts as

$$f^gW_{AB} = \sum_{ab} f(a)W_{AB}(a,b)g(b)
= \sum_{a_1b_1} f'(a_1)W_{AB}(0a_1,h(b_1))g'(b_1)
+ f''(a_1)W_{AB}(1a_1,h(b_1))g''(b_1)
+ f'(a_1)W_{AB}(0a_1,h(b_1))g'(b_1)
+ f''(a_1)W_{AB}(0a_1,h(b_1))g''(b_1).$$

Now observe the probability $W_{AB}(0a_1,h(b_1))$. By the definition of the function $h$ we can derive

$$W_{A_1B_1}(0,h(b_1)) = \sum_{b_1} W_{B_1}(b_1)W_{A_1B_1|B_1}(0,h(b_1))b_1 = p$$

and therefore, if we factor out the distribution of the first output pair $(a_1,b_1)$ we get

$$W_{AB}(0a_1,h(b_1)) = pW_{A_1B_1|A_1B_1}(a_1,b_1|0,h(b_1)).$$

For any $a_1, b_1$ the probability $W_{A_1B_1|A_1B_1}(a_1,b_1|0,h(b_1))$ is independent of the distribution of the first output pair. That means $P_1$ can as well be replaced by local circuitry. An alternative wiring distribution is obtained from the original wiring by fixing $a_1 = 0$ and $b_1 = h(b_1)$ for any $a_1, b_1$. Therefore, there exists a wiring distribution $W_{A_1B_1}$ on $n-1$ systems, for example the one described above, such that for each $a_1, b_1$ we have

$$W_{AB}(0a_1,h(b_1)) = pW_{A_1B_1}(a_1,b_1).$$

The above reasoning can be applied to all four parts of $f^gW_{AB}$ implying the existence of wiring distributions $W_{A_1B_1}^{(1)}, W_{A_1B_1}^{(2)}, W_{A_1B_1}^{(3)}$ and $W_{A_1B_1}^{(4)}$ such that

$$f^gW_{AB} = p \left( f^{g'}W_{A_1B_1}^{(1)} + f^{g''}W_{A_1B_1}^{(2)} \right) + \left( 1/2-p \right) \left( f^{g'}W_{A_1B_1}^{(3)} + f^{g''}W_{A_1B_1}^{(4)} \right).$$

Assume now the induction hypothesis that for any wiring distribution $W_{A_1B_1}$ on $n-1$ isotropic systems and functions $f,g : \{0,1\}^{n-1} \rightarrow \{0,1\}$ we have the upper bound

$$f^gW_{AB} \leq \delta_{n-1}^+(|f^{-1}(1)|,|g^{-1}(1)|).$$

Since the basis $f^gP_{AB} \leq \delta_{n-1}^+(|f^{-1}(1)|,|g^{-1}(1)|)$ trivially holds, we can conclude

$$f^gW_{AB} \leq p\delta_{n-1}^+(|f^{-1}(1)|,|g^{-1}(1)|)
+ (1/2-p)\delta_{n-1}^+(|f^{-1}(1)|,|g^{-1}(1)|)
\leq \delta_n^+(k,l).$$

We used that $0 \leq |f^{-1}(1)|,|g^{-1}(1)| \leq \min(k,2n-1)$ and $0 \leq |g^{-1}(1)|,|g^{-1}(1)| \leq \min(l,2n-1)$ holds. With the same arguments one can prove the lower bound $f^gW_{AB} \geq \delta_n^-(k,l)$ by induction.

The shown bounds to $f^gW_{AB}$ yield the following bound to the distillable nonlocality of isotropic systems:

**Corollary 1.** Let $P_{iso}$ be an isotropic system; then

$$D(n,P_{iso}) \leq \max_{0 \leq k^x,l^y \leq 2^n} \frac{1}{2} - \frac{(k^1 + l^0)}{2n-2} + 4\delta_n^+(k^0, l^0)
+ \delta_n^+(k^0, l^1) + \delta_n^+(k^1, l^0) - \delta_n^-(k, l^1).$$

where $\delta_n^+(k^x,l^y)$ and $\delta_n^-(k^x,l^y)$ are solutions to the problem instance $(P_{iso},n,k^x,l^y)$ for all $x, y$.

**Proof.** Note that from the definition of the inner product it follows that it is sufficient to consider positive values to find the maximum (see Lemma 3 in Appendix A).
Therefore we can omit the modulus in the CHSH value and get

\[
D(n, P) = \max_{0 \leq k, l \leq 2^n} \max_{DP} \text{NL}(DP) \\
= \max_{0 \leq k, l \leq 2^n} \max_{DP} \langle f_0, g_0 \rangle + \langle f_1, g_0 \rangle \\
+ \langle f_0, g_1 \rangle - \langle f_1, g_1 \rangle
\]

where the second maximum is over \( f_x \in C_k \) and \( g_y \in C_l \) for all \( x, y \). To complete the proof use \( \langle f_x, g_y \rangle = 1 - k^2/2^{n-1} - l^2/2^{n-1} + 4f_x^2W_{xy}g_y \) for all \( x, y \) and apply Lemma 1.

From the bound on distillable nonlocality of isotropic resources we derive a general bound as follows.

**Corollary 2.** For any \( P \), we have \( D(n, P) \leq D(n, P_{\text{iso}}) \) where \( P_{\text{iso}} \) shall be isotropic and minimize \( \text{NL}(P_{\text{iso}}) \) under the constraint \( P = qP_{\text{iso}} + P_L(1 - q) \) for any \( q \in [0, 1] \) and any local system \( P_L \).

**Proof.** By contradiction. Assume \( n P \)'s can be distilled to a value higher than \( D(n, P_{\text{iso}}) \). Given \( n P_{\text{iso}} \)'s we can construct a distillation protocol as follows: Combine each of the \( n P_{\text{iso}} \)'s with one local system \( P_L \) to get the \( n \) mixtures \( P = qP_{\text{iso}} + (1 - q)P_L \). Since we assumed that \( n P \)'s can be distilled above \( D(n, P_{\text{iso}}) \) we constructed a protocol on \( n P_{\text{iso}} \)'s that achieves the same.

Bounding \( D(n, P_{\text{iso}}) \) by Corollary 1 yields an exponential speedup compared to the naive search through all protocols. The idea is to use dynamic programming to solve the problem in Definition 4 in run-time \( 2^{O(n)} \) (Lemma 3 in Appendix B). Bounding \( D(n, P) \) by Corollary 2 requires finding the least nonlocal isotropic system \( P_{\text{iso}} \), such that \( P \) can be obtained from a convex combination of \( P_{\text{iso}} \) and a local system \( P_L \). This can be done efficiently using a linear program (see Lemma 4 in Appendix C). Finally we calculate a bound on \( D(n, P_{\text{iso}}) \) as described above to complete the bound to \( D(n, P) \).

To compare the results, an exhaustive search to determine \( D(n, P) \) exactly requires an effort of \( 2^{2^{O(n)}} \) an exponential increase to our solution.

**IV. APPLICATION**

Here, we illustrate Corollaries 1 and 2 on example applications. For this purpose consider a two-dimensional section in the binary nonsignaling polytope defined as the convex combination of a nonlocal vertex \( P_{NL} \) (a PR-box), an unbiased mixture of two local vertices \( P_{C} \) (perfectly correlated random bits) and the local isotropic system \( P_{L} \) (lying on the CHSH facet corresponding to \( P_{NL} \) [11, 18, 19]). We mix as:

\[
P_{\varepsilon, \delta} = \varepsilon P_{NL} + \delta P_{C} + (1 - \varepsilon - \delta)P_{L},
\]

where \( \varepsilon \in [0, 1], \delta \in [0, 1] \) and \( \varepsilon + \delta \leq 1 \) to ensure nonsignaling. By setting \( \delta = 0 \) we obtain a set of isotropic systems \( P_{\text{iso}} = P_{\varepsilon, 0} \), where \( \text{NL}(P_{\text{iso}}) = 2(\varepsilon + 1) \). These are special when it comes to distillation because any system can be turned into an isotropic system while preserving nonlocality (a procedure known as depolarization [24]). It is a long standing conjecture that isotropic systems cannot be distilled. Strong evidence to support this has been provided by Dukaric and Wolf [23], who showed that in the quantum region infinitely many examples of asymptotically undistillable isotropic systems must exist. Also, Short [22] gave a proof for two-copy impossibility, i.e., he showed that \( D(2, P_{\text{iso}}) = \text{NL}(P_{\text{iso}}) \) holds for any isotropic system. With the new bound at hand we can deliver additional evidence. We found that our isotropic bound (Corollary 1), calculated exactly for rational instances with Maple [26], confirms \( D(n^*, P_{\text{iso}}) = 2(\varepsilon + 1) = \text{NL}(P_{\text{iso}}) \) if walking on the isotropic line with step length \( \alpha (\varepsilon \in \{\alpha, 2\alpha, ..., 1 - \alpha\}) \) in all tested cases:

\[
\begin{array}{c|ccccccccc}
\alpha & 10^{-6} & 10^{-5} & 10^{-4} & 10^{-3} & 10^{-2} & 10^{-1} & 4^{-1} \\
n^* & 3 & 4 & 5 & 6 & 7 & 8 & 9 \\
\end{array}
\]

An immediate consequence is that in these settings our bound is tight, the optimal protocol simply imitates \( P_{\text{iso}} \). Interestingly, the bound on \( D(n, P_{\text{iso}}) \) drops significantly below \( \text{NL}(P_{\text{iso}}) \) if we restrict the protocol to use unbalanced functions, i.e., if we demand \( |f_0^{-1}(1)|, |f_1^{-1}(1)| \neq 2^{n-1} \) for example. Figure 1 illustrates the isotropic bound and this effect by an example calculation.

**FIG. 1.** Plotted are the calculated bounds to \( D(6, P_{1/5,0}) \) (z-axis) for \( 0 \leq k^0 + k^1 \leq 2^7 \) (x-axis) and \( 0 \leq l^0 + l^1 \leq 2^7 \) (y-axis). The peak in the center is the CHSH value of a fully balanced system of the same nonlocality as \( P_{1/5,0} \) sitting on the isotropic line.
the closer the system approaches the nonsignaling facet. we get the system shown to be limited by the bound to the weakest isotropic statement has direct implications in the general case. Therefore, we can use known asymptotic distillation bounds to generalize the above reasoning: Since \( D(\infty, P_{iso}) = NL(P_{iso}) \) holds for infinitely many isotropic systems in the quantum region [23], the existence of infinitely many sets, each with a different nonlocality limit, that are closed under wirings follows directly. This answers an open question posed in [20].

V. CONCLUSION

Distillable nonlocality as a measure for the nonlocality of a given set of correlations is sometimes preferable to the original CHSH nonlocality, which can be amplified by local operations and is therefore problematic in certain settings. Here, we derived bounds for the alternative measure. Bounding distillable nonlocality is useful if the bound can be calculated more efficiently than the exact value and is tight enough to be meaningful. We show both properties: An increased efficiency is achieved by a dynamic programming approach exploiting the recursive structure of a closely related optimization problem. The bound for isotropic systems (Corollary 1) is tight for the tested set of resources. It can be used to rule out distillation of up to 9 fixed copies supporting the long standing conjecture of asymptotic impossibility of distillation in the isotropic case. The second bound (Corollary 2) establishes an efficient and generic connection between the distillable nonlocality of isotropic systems and the general case. It explains the increasing distillation potential when departing from the isotropic line, that has been observed in many distillation protocols, and is tight for correlated nonlocal boxes.

The presented bounds are based on solutions to a recursive optimization problem that is new in this context. We believe that this abstraction can help to answer more general distillation questions and contributes to a deeper understanding of the possibilities of such protocols. It would be very interesting to find an efficient algorithm or an explicit solution or bound for the defined problem. This could be an important step towards proving general asymptotic (\( n \to \infty \)) bounds to distillable nonlocality in the future.

ACKNOWLEDGMENTS

This work was funded by the Swiss National Science Foundation (SNSF).

Appendix A: A symmetry of distillation protocols

Lemma 2. For any deterministic distillation protocol \( DP = (P^n, f_0, f_1, g_0, g_1) \) it holds that

\[
NL(P^n, f_0, f_1, g_0, g_1) = NL(P^n, \bar{f}_0, \bar{f}_1, g_0, g_1)
\]

FIG. 2. Here, we illustrate the idea behind the general distillation bound of Corollary 2 in a two-dimensional wedge. The distillable nonlocality of \( P_q \) (for \( q = 1/5, 2/5, 3/5, 4/5, 1 \)) is shown to be limited by the bound to the weakest isotropic system \( P_{iso} \), such that \( P_q \) can be expressed as a convex combination of \( P_{iso} \) and \( P_C \).

Another implication of the calculated isotropic bounds and Corollary 2 is the existence of many sets closed under \( (n < 10) \)-copy wirings. We have established the following: Given the set \( S(P_{iso}) \) of systems that are convex combinations of a certain \( P_{iso} \) and any local system. Then

\[
D(n, P_{iso}) = NL(P_{iso})
\]

implies the existence of a set \( S'(P_{iso}) \), where \( S(P_{iso}) \subseteq S'(P_{iso}) \), that is closed under \( n \)-copy wirings of systems in \( S(P_{iso}) \).

Note that the consequences of Corollary 2 are independent of how the isotropic bounds are derived. Since the corollary expresses a generic dependency between isotropic and general distillable nonlocality any isotropic statement has direct implications in the general case. Therefore, we can use known asymptotic
with \( f_\epsilon(a) = 1 - f(a) \) and \( g_\epsilon(a) = 1 - g(a) \) for all \( a \in \{0, 1\}^n \) and all \( x, y \in \{0, 1\} \).

\( \)Proof.\( \) Observe that for any \( f, g \in \{0, 1\}^n \) and any wiring distribution \( W_{AB} \) on \( n \) systems we have
\[
(f, g) = (1 - 2f)^T W_{AB} (1 - 2g) \\
= -((1 - 2f)^T) W_{AB} (1 - 2g) \\
= -(\delta, g).
\]
Apply it on
\[
NL(DP) = |\langle f_0, g_0 \rangle + \langle f_0, g_1 \rangle + \langle f_1, g_0 \rangle - \langle f_1, g_1 \rangle|
\]
to prove the lemma. \( \square \)

Appendix B: Calculate \( \delta_n^+ \) and \( \delta_n^- \) by dynamic programming

Lemma 3. Given numbers \( n > 1 \) and \( 0 \leq k, l \leq 2^n \), the values \( \delta_n^+(k, l) \) and \( \delta_n^-(k, l) \) can be calculated in \( 2^{O(n)} \) steps.

\( \)Proof.\( \) Taking advantage of the recursive structure of \( \delta_n^+ \) and \( \delta_n^- \) we build the two tables
\[
D_n^+ = \{\delta_n^+(i, j)\}_{0 \leq i, j \leq 2^n} \quad \text{and} \quad D_n^- = \{\delta_n^-(i, j)\}_{0 \leq i, j \leq 2^n}
\]
recursively from the tables \( D_{n-1}^+ \) and \( D_{n-1}^- \). By reading out from \( D_{n-1}^+ \) and \( D_{n-1}^- \) the best of all possible solutions on \( (i, j) \) for every entry \( D_n^+(i, j) = \delta_n^+(i, j) \) and \( D_n^-(i, j) = \delta_n^-(i, j) \) we fill the tables \( D_n^+ \) and \( D_n^- \) element-wise. To save time some simple reduction rules derivable from the inner product definition and the trivial symmetries \( \delta_n^+(i, j) = \delta_n^+(j, i) \) and \( \delta_n^-(i, j) = \delta_n^-(j, i) \) for each \( i, j \) can be used here – only an eighth of the table must actually be calculated. The base tables are given by \( D_0^+(k, l) = 1 \) and \( D_0^-(k, l) = 1 \). Once the tables are filled, for fixed \( n \) and \( P \) the value \( D(n, P) \) can be bounded with the data in \( D_n^+ \) and \( D_n^- \) by iterating over all possible Boolean function classes \( 0 \leq |f^{(1)}| \leq 2^n \) and \( 0 \leq |g^{(1)}| \leq 2^n \). It requires roughly \( \sum_{i=2}^{2n-5} 2^{\delta_i} \in 2^{O(n)} \) steps to fill the tables in this way – a workload that is feasible for \( n < 10 \) on a normal PC. \( \square \)

Appendix C: Finding the optimal \( P_\text{iso} \)

Bounding \( D(n, P) \) by Corollary 2 requires finding the least nonlocal isotropic system \( P_\text{iso} \) such that \( P \) can be obtained from a convex combination of \( P_\text{iso} \) and a local system \( P_L \).

Lemma 4. Given a system \( P \) one can efficiently determine the isotropic system \( P_\text{iso} \), such that \( NL(P_\text{iso}) \) is minimal under the constraint \( P = qP_\text{iso} + P_L(1 - q) \) for any \( q \in [0, 1] \) and any local system \( P_L \).

\( \)Proof.\( \) The system \( P \) identifies a particular CHSH inequality, now called CHSH\(_P\), and therefore a family of isotropic systems denoted
\[
P_\text{iso}(\epsilon) = \epsilon P_NL + (1 - \epsilon) P_F
\]
with \( \epsilon \in [0, 1] \) and \( P_NL \) as the nonlocal vertex corresponding to CHSH\(_P\) and the unique isotropic system \( P_F \in \text{CHSH}_P \). To find the optimal \( \epsilon \) we first calculate the local part of \( P \) \( \text{[14]} \) with the linear program defined by Fitzi et al. \( \text{[13]} \). The local part of \( P \) is the optimal value of the following linear program:
\[
\text{max} \sum_i p_i \\
\text{subject to} \sum_i p_i P_{L,i} \leq P_{AB}(a, b) \quad p_i \geq 0.
\]
Here, \( P_{L,i} \) denote the vertices of the local polytope. Since \( \sum_i p_i \) is maximal and there is only one nonlocal vertex violating CHSH\(_P\) we have
\[
P = \sum_i p_i P_{L,i} + (1 - \sum_i p_i) P_{NL}.
\]
Now, we decompose the local system \( P^* = \sum_j \frac{p_j}{\sum_i p_i} P_{L,j} \) into
\[
P^* = p_f P_F + (1 - p_f) P_L
\]
where \( P_L \) is any local system. By the local version of Lemma 1 in \( \text{[15]} \) (holds since normalization and locality are linear properties) the maximal \( p_f \) possible is given by
\[
p_f = \min_{a, b, x, y} \frac{P_{AB}(a, b)}{P_{AB}(a, y)}.
\]
Therefore, the parameter \( \epsilon \) fixing the optimal \( P_\text{iso}(\epsilon) \) can easily be constructed via
\[
P = \sum_i p_i P^* + (1 - \sum_i p_i) P_{NL} \\
= \sum_i p_i (p_f P_F + (1 - p_f) P_L) + (1 - \sum_i p_i) P_{NL} \\
= \sum_i p_i (1 - p_f) P_L + \sum_i p_f p_f P_F + (1 - \sum_i p_i) P_{NL} \\
= \sum_i p_i (1 - p_f) P_L + (1 - \sum_i p_i (1 - p_f)) P_{iso}(\epsilon),
\]
with
\[
\epsilon = \frac{1 - \sum_i p_i}{\sum_i p_i p_f + 1 - \sum_i p_i}.
\]
So, \( \epsilon \) can be determined efficiently. \( \square \)
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