Communications in Mathematical Physics



Universal Gaps for XOR Games from Estimates on Tensor Norm Ratios

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Received: 26 November 2018 / Accepted: 8 December 2019 Published online: 7 March 2020 – © The Author(s) 2020

Abstract: We define and study XOR games in the framework of general probabilistic theories, which encompasses all physical models whose predictive power obeys minimal requirements. The bias of an XOR game under local or global strategies is shown to be given by a certain injective or projective tensor norm, respectively. The intrinsic (i.e. model-independent) advantage of global over local strategies is thus connected to a universal function r(n, m) called 'projective-injective ratio'. This is defined as the minimal constant ρ such that $\|\cdot\|_{X\otimes_{\pi}Y} \leq \rho \|\cdot\|_{X\otimes_{\varepsilon}Y}$ holds for all Banach spaces of dimensions dim X = n and dim Y = m, where $X \otimes_{\pi} Y$ and $X \otimes_{\varepsilon} Y$ are the projective and injective tensor products. By requiring that X = Y, one obtains a symmetrised version of the above ratio, denoted by $r_s(n)$. We prove that $r(n, m) \ge 19/18$ for all $n, m \ge 2$, implying that injective and projective tensor products are never isometric. We then study the asymptotic behaviour of r(n, m) and $r_s(n)$, showing that, up to log factors: $r_s(n)$ is of the order \sqrt{n} (which is sharp); r(n, n) is at least of the order $n^{1/6}$; and r(n, m) grows at least as min $\{n, m\}^{1/8}$. These results constitute our main contribution to the theory of tensor norms. In our proof, a crucial role is played by an $\ell_1/\ell_2/\ell_{\infty}$ trichotomy theorem' based on ideas by Pisier, Rudelson, Szarek, and Tomczak-Jaegermann. The main operational consequence we draw is that there is a universal gap between local and global strategies in general XOR games, and that this grows as a power of the minimal local dimension. In the quantum case, we are able to determine this gap up to universal constants. As a corollary, we obtain an improved bound on the scaling of the maximal quantum data hiding efficiency against local measurements.

1. Introduction

One of the most prominent conceptual contributions of the celebrated 1964 paper by Bell [1] is to point out that the implications of the quantum mechanical predictions extend far beyond the very same formalism that is used to deduce them, and shed light on some of the deepest secrets of Nature. The scenario considered by Bell features two distant parties who share a quantum entangled state and make local quantum measurements on it. The main contribution of [1] is to show that the resulting correlations cannot be explained by any local 'hidden variable' theory. Here we want to stress that the argument does not depend on the correctness of quantum mechanics as the ultimate theory of Nature, but rather only on the accuracy of its predictions concerning the above experimental setting. In other words, any alternative theory that leads to the same predictions in the same setting will also be subjected to Bell's theorem. In this spirit, we deem it important to understand what features of information processing in composite systems are truly *intrinsic*, meaning that they are common to all conceivable physical theories.

A suitable way to formalise the concept of a physical theory in this context is provided by the mathematical machinery of general probabilistic theories (GPTs) [2–4]. It is sometimes convenient to think of GPTs as generalisations of finite-dimensional quantum mechanics, where the set of unnormalised states is not assumed to be the cone of positive semidefinite matrices, but it is rather taken to be an arbitrary convex cone in a finitedimensional real vector space. As is well known, a GPT makes the host vector space a Banach space in a canonical way, by equipping it with a so-called base norm.

The starting point of our investigation is the study of XOR games in the rich landscape of GPTs. We remind the reader that any XOR game can be equivalently cast in terms of state discrimination queries subjected to locality constraints, so that our analysis applies equally well to these problems. XOR games are arguably the simplest examples of two-prover one-round games and feature two cooperating players, Alice and Bob, and a third party known as the referee. The referee asks the players some 'questions' by sending them states of some physical system modelled by a GPT. The correctness of the one-bit answers the players provide upon measuring the state depends only on their parity. According to whether Alice and Bob are allowed to carry out product or global measurements, one talks about local or global strategies (see the two paragraphs right after Definition 3 for the precise definition of these strategies). In general, the winning probability can be significantly larger in the latter than in the former case.

While the quantitative details of this phenomenon will in general depend on the particular physical system modelling the questions, our work is instead motivated by the wish to understand which behaviours are universal, and thus pertain to the intrinsic nature of XOR games. This line of investigation brings us to develop an extensive connection with the theory of tensor norms on finite-dimensional Banach spaces, which has already proved to be instrumental in the study of classical and quantum XOR games [5]. While in these more standard settings one deals with *specific examples* of tensor norms, the analysis of games played over arbitrary GPT models requires a *systematic* understanding of *general* tensor norms.

The main problem we investigate here asks for the maximal gap that can be guaranteed to exist between the winning probabilities associated with global and local strategies in XOR games played over GPTs of fixed local dimensions n and m. In analogy with the classical case, we show that such winning probabilities are given by simple expressions involving respectively the projective and injective tensor norms induced by the local GPTs through their native Banach space structures. Comparing them in a modelindependent fashion prompts us to investigate the least constant of domination of the injective over the projective tensor norm over all pairs of normed spaces of fixed dimensions. We call this function 'projective/injective ratio', or ' π/ε ratio' for short. When seen from the point of view of pure mathematics, this universal function of (n, m) encodes some information regarding Grothendieck's theory of tensor products of Banach spaces. At the same time, the operational interpretation we construct here guarantees that the same object captures some intrinsic feature of general XOR games.

Our main result is that the π/ε ratio associated with two *n*-dimensional Banach spaces scales at least as $n^{1/6}$ (up to logarithmic factors), implying that global strategies for XOR games are intrinsically much more effective than local ones in a precise asymptotic sense. We ask the question whether this scaling can be improved up to $n^{1/2}$, and bolster this hypothesis by showing that it holds true (again, up to log factors) when two copies of the same space are considered. Interestingly, this question is intimately connected to the problem of estimating the radius of the weak Banach-Mazur compactum, which has also been conjectured to be of order $n^{1/2}$ [6,7]. We also consider the problem of establishing dimension-independent lower bounds for the π/ε ratio. We prove that for all pairs of Banach spaces X, Y of dimension at least 2 (and possibly infinite), there is a nonzero tensor in $X \otimes Y$ whose projective norm is at least 19/18 times its injective norm. In particular, these norms are always different, which seems to be a new observation. This should be compared with the famous construction by Pisier [8] of an infinite-dimensional Banach space X such that the injective and projective norms on $X \otimes X$ are equivalent. Finally, we solve the problem of computing the π/ε ratio for some specific examples of physically relevant Banach spaces. Most notably, we establish that it is of the order $\min\{n, m\}^{3/2}$ for $S_1^{n, \text{sa}} \otimes S_1^{m, \text{sa}}$, where $S_1^{k, \text{sa}}$ stands for the space of $k \times k$ Hermitian matrices endowed with the trace norm. The importance of this special case stems from the fact that $S_1^{k,sa}$ is the natural Banach space associated with a k-level quantum system.

The rest of the paper is structured as follows. Throughout this section, we provide very brief introductions to the GPT formalism (Sect. 1.1), to the theory of tensor norms (Sect. 1.2), and to XOR games (Sect. 1.3). In Sect. 2, we state our main results and broadly discuss some of the proof techniques we developed. Section 3 presents some general properties of the π/ε ratio, connecting it with other concepts in functional analysis. There, we find the universal lower bound of 19/18, and solve the quantum mechanical case up to multiplicative constants. Section 4 deals with the problem of determining the asymptotic scaling of the π/ε ratio, either for two copies of the same space, or in the fully general setting of two normed spaces of different dimensions. In order to improve the accessibility of the paper we added several Appendices, where some extra information can be found. "Appendix A" investigates how our operational interpretation of the injective tensor norm is affected by the introduction of a bounded amount of two-way communication, while "Appendix B" provides a proof of the useful fact that any normed space is 2-isomorphic to a base norm space. Finally, "Appendix C" gathers the functional-analytic background that is used throughout the paper and which may be unfamiliar to a non-specialist reader. In particular, we sketch there proofs of various statements which can be deduced from known results, but some elucidation is needed for non-experts.

1.1. General probabilistic theories. The origins of the formalism of general probabilistic theories (GPTs) lie in the attempt to axiomatise quantum mechanics, rebuilding it upon operationally motivated postulates rather than upon more evasive concepts such as 'wave function' and 'microscopic system'. Although these ideas can be found already in

some antecedent works [9,10], the first major contributions were made by the 'Marburg school' led by Ludwig [11,12]. This resulted in an intense debate around the nascent GPT formalism, which took place in a series of papers published in Communications in Mathematical Physics [13–18]. For an account of the early development of the field, we refer the interested reader to [2]. A more modern point of view can be found in [4, Chapter 1]. This foundationally motivated interest has seen a revival in the last two decades, with much effort being focused on attempts to 'reconstruct' quantum mechanics starting from first principles [19–23]. At the same time, GPTs have become central to quantum information science, as they provide indispensable tools to analyse information processing beyond classical theories, see for instance [24–32]. An introduction to the GPT framework that will suffice for our purposes can be found in [4, Chapter 2] (see also [32, Section 2]). Throughout this subsection, we limit ourselves to recalling the basics and to fixing the notation.

Definition 1. A general probabilistic theory is a triple (V, C, u), where: (i) *V* is a finitedimensional real vector space; (ii) $C \subset V$ is a closed, convex, salient and generating cone; and (iii) *u*, called the **order unit** or the **unit effect**, is a functional in the interior of the dual cone $C^* := \{x^* \in V^* : x^*(x) \ge 0 \forall x \in C\}$. GPTs will be denoted by capital letters such as *A*, *B* etc., which – with a slight abuse of notation – identify also the underlying physical systems. We call dim *V* the **dimension** of the GPT.

On the mathematical level, we can think of (V, C) as an ordered vector space, the ordering being given by $x \leq y \Leftrightarrow y - x \in C$. Also the dual vector space can be thought of as ordered by the dual cone C^* . In this language, the functional u is said to be strictly positive, since $x \ge 0$ and u(x) = 0 implies x = 0. The states of the physical system modelled by (V, C, u) are represented by vectors in $C \cap u^{-1}(1) =: \Omega$. The compact convex set Ω is called the **state space** of the GPT, and accordingly we will sometimes refer to C as the cone of unnormalised states. Convexity here plays an operationally relevant role, as the process of preparing a system in a state ω_0 with probability p and ω_1 with probability 1 - p, and later forgetting the value of the binary random variable associated with its preparation, leaves the system in the state $p\omega_0 + (1 - p)\omega_1$.

The GPT formalism allows us to make probabilistic predictions of the outcomes of measurements performed on a certain state. In this context, a **measurement** is a finite collection $(e_i)_{i \in I}$ of functionals in the order interval [0, u] (generically called **effects**) that add up to the order unit, i.e. such that $\sum_i e_i = u$ (**normalisation**). The probability of obtaining the outcome *i* upon measuring the state $\omega \in \Omega$ is evaluated as $e_i(\omega)$. Throughout this paper, we will always make the so-called *no-restriction hypothesis*, which guarantees that all normalised collections of effects identify a physically legitimate measurement [33]. We denote by **M** the set of all measurements associated with a certain GPT, adding a subscript to identify it if needed.

Equipping an ordered vector space (V, C) with a GPT structure entails selecting a special positive functional on it, i.e. the unit effect u. In turn, this special functional can be used to define a norm on the dual space V^* . By definition, the unit ball of this norm is the **order interval** [-u, u], and for $x^* \in V^*$ one has

$$\|x^*\| := \min\left\{t \ge 0 : \ x^* \in t[-u, u]\right\}.$$
(1)

This choice makes V^* a so-called **order unit space** [34,35]. The corresponding Banach space structure induced on V is that of a **base norm space** [36]. The norm on V is given by any of the two expressions

$$\|x\| = \max_{x^* \in [-u,u]} |x^*(x)|$$
(2)

$$= \min \left\{ u(x_{+} + x_{-}) : x = x_{+} - x_{-}, x_{\pm} \ge 0 \right\}.$$
(3)

The equivalence is an easy consequence of the strong duality of conic optimisation programs [37]; alternatively, it can be established by checking that the convex body $K := \operatorname{conv}(\Omega \cup -\Omega) \subset V$ is the unit ball for the base norm, while $||x^*|| = \sup\{|x^*(x)| : x \in K\}$ for any $x^* \in V^*$.

This more or less exhausts the description of single systems within the GPT framework. Note that dynamics is not part of this very basic picture, which is limited to socalled 'prepare-and-measure procedures'. Time evolution can be accounted for within this formalism, but this goes beyond the scope of the present paper. What we will need here is instead the extension of the formalism to the case of composite systems. We will be mainly concerned with the simplest case of a bipartite system formed by two subsystems A and B described by local GPTs (V_A, C_A, u_A) and (V_B, C_B, u_B) . The theory modelling the composite AB will be denoted by $AB = (V_{AB}, C_{AB}, u_{AB})$. Under very reasonable assumptions [38,39], the main one being that bipartite states are always uniquely determined by the statistics produced by local measurements (a principle that goes under the name of **local tomography**), one can identify V_{AB} with the tensor product of the local vector spaces, i.e. $V_{AB} \simeq V_A \otimes V_B$. When this identification is made, one obtains also $u_{AB} = u_A \otimes u_B$. To fully specify the joint system, one still needs to identify the cone of unnormalised states C_{AB} . It turns out that such a choice cannot be made a priori on the ground of some indisputable axiom, but has to be based on some information regarding the actual physics of the system. However, the operational interpretation of the theory puts some nontrivial constraints on C_{AB} , in the form of a lower and upper bound with respect to the inclusion relation. Namely, we have

$$C_A \underset{\min}{\otimes} C_B \subseteq C_{AB} \subseteq C_A \underset{\max}{\otimes} C_B , \qquad (4)$$

where

$$C_A \bigotimes_{\min} C_B := \operatorname{conv} \left(C_A \otimes C_B \right), \tag{5}$$

$$C_A \underset{\max}{\otimes} C_B := \left(C_A^* \underset{\min}{\otimes} C_B^* \right)^*.$$
(6)

The two constructions (5) and (6) are called **minimal** and **maximal tensor product**, respectively. In (4), the lower bound comes from the fact that any tensor product of local states must represent a valid state, while – dually – the fact that any tensor product of local effects must be an effect of the joint system leads to the upper bound. In (5) we used the notation $C_A \otimes C_B := \{x \otimes y : x \in C_A, y \in C_B\}$. In what follows, we will call **admissible** any composite *AB* whose associated cone C_{AB} satisfies (4). Also, we will denote by $A \underset{\text{min}}{\otimes} B$ and $A \underset{\text{max}}{\otimes} B$ those corresponding to the choices (5) and (6) for C_{AB} .

We conclude this brief presentation of the GPT formalism by discussing the two physically most relevant examples, i.e. classical probability theory and quantum mechanics. Classical probability theory can be viewed as the GPT

$$\operatorname{Cl}_{d} := \left(\mathbf{R}^{d}, \mathbf{R}_{+}^{d}, u\right),\tag{7}$$

where \mathbf{R}^{d}_{+} is the cone of entrywise non-negative vectors, and $u(x) := \sum_{i=1}^{d} x_{i}$ for all $x \in \mathbf{R}^{d}$. The induced base norm coincides with the ℓ_{1} -norm $||x||_{\ell_{1}} := \sum_{i=1}^{d} |x_{i}|$.

Composing classical systems is easy, for when either C_A or C_B is simplicial (i.e. a linear image of \mathbf{R}^4_{\perp}) minimal and maximal tensor product coincide.

An *n*-level quantum mechanical system is modelled by the GPT

$$QM_n := (\mathsf{M}_n^{\mathrm{sa}}, \mathrm{PSD}_n, \mathrm{tr}), \qquad (8)$$

where M_n^{sa} is the space of $n \times n$ Hermitian matrices, PSD_n the cone of positive semidefinite matrices, and tr represents the trace functional. The quantum mechanical base norm is the appropriate non-commutative generalisation of the ℓ_1 -norm, i.e. the trace norm $||X||_1 := \text{tr } \sqrt{X^{\dagger}X}$. The base norm space of $n \times n$ Hermitian matrices endowed with the trace norm will be denoted by $S_1^{n,sa}$. In contrast with the classical case, for quantum mechanical composition rules become an issue. In fact, the standard quantum mechanical composition rule dictates that if $A = QM_n$ and $B = QM_m$ then $AB = QM_{nm}$. The corresponding cone C_{AB} is well known to make both inclusions in (4) strict: provided $n, m \ge 2$, we have

$$\operatorname{PSD}_{n} \underset{\min}{\otimes} \operatorname{PSD}_{m} \subsetneq \operatorname{PSD}_{nm} \subsetneq \operatorname{PSD}_{n} \underset{\max}{\otimes} \operatorname{PSD}_{m}.$$
(9)

Indeed, the left member of (9) is the cone of separable operatorss, and thus the strictness of the left inclusion reflects the existence of entanglement. Since the positive semidefinite cone is selfdual, it follows by duality that the right inclusion is also strict. Note that the right member of (9) is the cone of the so-called block-positive operators, which can serve as entanglement witnesses.

1.2. Tensor norms. We start by recalling the basic theory of tensor products of normed spaces. In what follows $B_X := \{x \in X : ||x|| \le 1\}$ will denote the unit ball of a Banach space *X*. There are at least two canonical ways in which one can construct a norm on a generic tensor product $X \otimes Y$ of finite-dimensional real Banach spaces [40–42]. The **injective norm** of a tensor $z \in X \otimes Y$ is defined by the expression

$$||z||_{X\otimes_{\varepsilon}Y} := \max\left\{ (x^* \otimes y^*)(z) : x^* \in B_{X^*}, y^* \in B_{Y^*} \right\},\tag{10}$$

while its projective norm is given by

$$\|z\|_{X\otimes_{\pi}Y} := \min\left\{\sum_{i} \|x_i\| \|y_i\| : z = \sum_{i} x_i \otimes y_i\right\}.$$
 (11)

In Sect. 1.1, we learnt that the vector space associated with a GPT carries a natural norm, i.e. the base norm given by (3). Since a joint system *AB* lives on the tensor product $V_A \otimes V_B$ of the local vector spaces, it is natural to ask whether either of the above tensor norms admits an operational interpretation in this context. Indeed, it turns out that [32, Proposition 22]

$$\|\cdot\|_{AB} \leqslant \|\cdot\|_{A \underset{n}{\otimes} B} = \|\cdot\|_{V_A \otimes_\pi V_B}.$$

$$(12)$$

for all admissible composites AB. The last equality tells us that the projective norm corresponds to the base norm associated with the minimal tensor product of the two theories.

One may thus be led to conjecture that an analogous identity exists between $\|\cdot\|_{V_A \otimes_{\varepsilon} V_B}$ and $\|\cdot\|_{A \otimes_{\varepsilon} B}$, but the example of two classical probability theories reveals that this is not the case. We will find an adequate operational interpretation for the injective tensor norm in the forthcoming Sect. 1.3. The most elementary properties of injective and projective norms is perhaps the inequality

$$\|\cdot\|_{X\otimes_{\varepsilon}Y} \leqslant \|\cdot\|_{X\otimes_{\pi}Y}, \tag{13}$$

valid for all *X*, *Y*. Moreover, since the space $X \otimes Y$ is of finite dimension, these two norms will always be equivalent, i.e. there will exist a constant $1 \leq C < \infty$ such that

$$\|\cdot\|_{X\otimes_{\varepsilon}Y} \leqslant \|\cdot\|_{X\otimes_{\pi}Y} \leqslant C\|\cdot\|_{X\otimes_{\varepsilon}Y}.$$
(14)

Denote by $\rho(X, Y)$ the smallest constant *C* satisfying this inequality. It is straightforward to verify that it is formally given by the following optimisation:

$$\rho(X,Y) := \sup_{0 \neq z \in X \otimes Y} \frac{\|z\|_{X \otimes_{\pi} Y}}{\|z\|_{X \otimes_{\varepsilon} Y}}.$$
(15)

For reasons that will soon become clear, in this paper we are interested in studying the range of values of the function $\rho(X, Y)$ across all pairs of spaces of fixed dimensions.

Definition 2. The **projective/injective ratio**, or π/ε ratio for short, is the following universal function over pairs of integers $n, m \ge 2$:

$$r(n,m) := \inf_{\substack{\dim X=n \\ \dim Y=m}} \rho(X,Y),$$
(16)

where the optimisation is understood to be over all pairs of finite-dimensional Banach spaces X, Y of fixed dimensions n, m. A slight modification of the above function (16) yields the **symmetric projective/injective ratio**:

$$r_s(n) := \inf_{\dim X = n} \rho(X, X), \qquad (17)$$

where $n \ge 2$ and the infimum is taken over all Banach spaces of dimension *n*.

One could equally well investigate analogous quantities where the infimum in the above optimisations is replaced by a supremum, however it turns out that these can be evaluated exactly. In fact, it has been shown that [32, Proposition 21]

$$R(n,m) := \sup_{\substack{\dim X=n\\\dim Y=m}} \rho(X,Y) = \min\{n,m\}.$$
(18)

In light of this, in the rest of the paper we shall be concerned with the π/ε ratios as constructed in Definition 2. By considering the examples $X = \ell_1^n$ and $Y = \ell_2^m$ in (16) (and assuming without loss of generality that $n \leq m$), one can see that

$$1 \leqslant r(n,m) \leqslant \sqrt{\min\{n,m\}} \quad \forall n,m.$$
⁽¹⁹⁾

For an explicit proof, see the discussion preceding (59). To upper bound the symmetrised ratio one can consider two copies of ℓ_1^n , which yields the slightly worse estimate [32, Example 29] (we compute a sharper upper bound on $\rho(\ell_1^n, \ell_1^n)$, which is equivalent to $\sqrt{\pi/2}\sqrt{n}$ as *n* tends to infinity, in Proposition 14)

$$1 \leqslant r_s(n) \leqslant \rho(\ell_1^n, \ell_1^n) \leqslant \sqrt{2n} \quad \forall n.$$
⁽²⁰⁾

Note that although $r(n, n) \leq r_s(n)$ for all n, it may conceivably happen that $r(n, n) < r_s(n)$. In other words, it is possible that the infimum in (16) is not achieved on two copies

of the same space even when n = m. However, we do not know this for a fact, even when n = 2 [cf. (77), (78)].

The above inequalities exhaust the elementary properties of the π/ε ratios, and leave open many interesting questions, whose thorough investigation constitutes our main contribution. For a summary of the results we obtain on these quantities, we refer the reader to Sect. 2.

1.3. XOR games. A simple but extremely useful setting where different physical models can be studied and compared from the point of view of information processing is that defined by XOR games. In these games, a referee interacts with two players Alice and Bob, who can cooperate with each other in order to maximise their winning probability. In the classical setting, the referee chooses a pair of questions according to a publicly known distribution and sends one question to each player. Then, the players are requested to provide a one-bit answer each, and the winning condition of the game, for a given pair of questions, only depends on the parity of the answers. In the basic local setting, the players can agree in advance on a strategy for their answers but they are not allowed to communicate with each other once the game has started.

These games are arguably central in theoretical computer science, mainly because of their simplicity and broad applicability to different topics such as interactive proof systems, hardness of approximation, and the PCP theorem. In addition, XOR games have played a major role in quantum information theory since they were first considered in [43]. In fact, these games had already been implicitly considered in the context of the study of quantum nonlocality [1,44], by means of their equivalent formulation in terms of correlation Bell inequalities. Their systematic study was initiated by Tsirelson [45]. Far from being purely theoretical objects, in the last years these games have been crucial in the development of device-independent quantum cryptography and random numbers generators.

Motivated by their relevance for theory and applications, and drawing from previous works that put forth suitable quantum generalisations [46,47], in this paper we introduce XOR games in the context of GPTs. In this more general setting, the two players' system will be described by some bipartite GPT $AB = (V_A \otimes V_B, C_{AB}, u_A \otimes u_B)$. The referee samples the questions from a finite alphabet *I*, the probability of drawing *i* being denoted by p_i . The answers are represented by a collection of bits $(c_i)_{i \in I} \in \{0, 1\}^{|I|}$, while the questions are described by states $\omega_i \in C_{AB} \subset V_A \otimes V_B$. Upon being asked the question ω_i , the players output answers $a \in \{0, 1\}$ and $b \in \{0, 1\}$, respectively, and the winning condition takes the form $a \oplus b = c_i$. The players' behaviour can be modelled by a suitable measurement $M = (g_{ab})_{ab \in \{00,01,10,11\}} \in \mathbf{M}_{AB}$ over AB, with $g_{ab}(\omega_i)$ representing the probability that the answers a, b are given when the question ω_i has been asked. We can then formalise the notion of XOR game as follows.

Definition 3. An **XOR game** *G* is a quadruple (*AB*, ω , *p*, *c*), where: (i) *AB* is a bipartite GPT; (ii) $\omega = (\omega_i)_{i \in I}$ is a discrete collection of states over *AB*; (iii) *p* is a probability distribution over the set *I*; and (iv) $c = (c_i)_{i \in I}$ is a set of bits. A **strategy** for the players is a measurement $M = (g_{ab})_{ab \in \{00,01,10,11\}}$ over *AB*.

Before we delve into the study of XOR games over GPTs, let us point out some caveats in the terminology. According to the standard conventions, a **classical XOR** game is more than an XOR game played over the composite $AB = Cl_{nm}$ formed by two classical GPTs $A = Cl_n$ and $B = Cl_m$ defined by (7). In fact, it is usually understood that in this case the questions are taken from the standard basis of $V_{AB} = \mathbb{R}^{nm}$, i.e.

 $\omega_{xy} := v_x \otimes v_y$, where $1 \leq x \leq n$ and $1 \leq y \leq m$ are integers, and v_k is the *k*-th vector in the standard basis of \mathbf{R}^k . In view of the perfect local distinguishability of the questions, one usually refers directly to the labels *x* and *y* as the questions. In compliance with the established conventions, from now on we will stick to the above definition of a classical XOR game.

The prototypical (and simplest) example of a strategy, called a **local strategy**, consists in the players performing a product measurement $(e_a \otimes f_b)_{a,b \in \{0,1\}} \in \mathbf{M}_A \otimes \mathbf{M}_B$. The opposite case is naturally that of a **global strategy**, corresponding to the case of Alice and Bob having access to global measurements $(g_{ab})_{a,b \in \{0,1\}} \in \mathbf{M}_{AB}$, but one can equally well consider some intermediate scenarios where the players are allowed a bounded amount of communication before they are required to output the answers. Some of these variations on the theme are examined in "Appendix A". From the above picture it follows that XOR games can be equivalently formulated as instances of state discrimination problems, possibly subjected to some special constraints dictated by locality.

One is usually interested in maximising the success probability of the players given a certain set of measurements they have access to. Since an XOR game can always be won with probability 1/2 by just answering randomly, it is standard to quantify the effectiveness of the players' strategy by introducing the **bias** $\beta(G)$ of the game G:

$$\beta(G) := P_{\text{winning}}(G) - P_{\text{losing}}(G) = 2P_{\text{winning}}(G) - 1.$$
(21)

It is understood that the bias depends also on the strategy adopted by the players, which we specify with a subscript. The following result yields explicit expressions for the bias corresponding to local and global strategies.

Theorem 1. Consider an XOR game $G = (AB, \omega, p, c)$, where $AB = (V_A \otimes V_B, C_{AB}, u_A \otimes u_B)$. Define $z_G := \sum_{i \in I} p_i (-1)^{c_i} \omega_i \in V_A \otimes V_B$. Then the biases corresponding to local and global strategies evaluate to

$$\beta_{\text{LO}}(G) = \|z_G\|_{V_A \otimes_{\varepsilon} V_B} , \qquad (22)$$

$$\beta_{\text{ALL}}(G) = \|z_G\|_{AB} , \qquad (23)$$

respectively. Here, $\|\cdot\|_{AB}$ is the base norm associated with the GPT AB. In particular, we obtain $\beta_{ALL}(G) \leq \|z_G\|_{V_A \otimes_{\pi} V_B}$ for all composites AB, and the bound is achieved when $AB = A \bigotimes_{m_B} B$.

Proof. It is not difficult to realise that when the players adopt global strategies, their task is equivalent to a state discrimination problem with all measurements on the composite system being available. Hence, (23) follows from [32, Lemma 7]. As for (22), note that a pair of local measurements ($e_a \otimes f_b$)_{a,b \in \{0,1\}} yields

$$P_{\text{winning}}(G) - P_{\text{losing}}(G)$$

$$= \sum_{i} p_{i} \left(\sum_{a,b: a \oplus b = c_{i}} (e_{a} \otimes f_{b})(\omega_{i}) - \sum_{a,b: a \oplus b \neq c_{i}} (e_{a} \otimes f_{b})(\omega_{i}) \right)$$

$$= \sum_{i} p_{i} \sum_{a,b} (-1)^{a+b+c_{i}} (e_{a} \otimes f_{b})(\omega_{i})$$

$$= \left(\left(\sum_{a} (-1)^{a} e_{a} \right) \otimes \left(\sum_{b} (-1)^{b} f_{b} \right) \right) (z_{G}).$$

That the optimisation over all local measurements yields (22) is a consequence of the elementary fact that

$$\{e_0 - e_1 : e_0 \in [0, u_A], e_0 + e_1 = u_A\} = [-u_A, u_A],$$
(24)

and analogously for system *B*. The last claim follows from (12). \Box

Remark. The value of the bias under local strategies depends only on the local structure of the GPTs *A* and *B*, and is thus independent of the particular rule we chose to compose them.

Theorem 1 generalises the mathematical description of the non-entangled bias of classical and quantum XOR games [5,47], and yields the operational interpretation of the injective norm we were seeking. In "Appendix A" we show that this interpretation is 'robust', in the sense that even allowing the players a bounded amount of communication before they output the answers does not increase the bias by more than a factor equal to the product of the dimensions of the GPTs used to carry the messages. The same type of argument also shows that not much can be gained if the players have access to a pre-shared physical system of bounded dimension.

For classical XOR games more is true: namely, Tsirelson's theorem [45] states that even assistance by entangled quantum states of *arbitrarily large* dimension does not allow for a significant improvement over product strategies. However, this has to be regarded as a peculiar feature of quantum systems, deeply linked with the underlying Hilbert space structure through Grothendieck's inequality. It is therefore not surprising that it does not carry over to the general GPT setting. In fact, it turns out that a classical XOR game can always be won with probability 1 if one allows for assistance from a well-chosen set of non-signalling correlations¹ (we point that non-signalling correlations can be viewed as GPTs [4, Section 2.3.4]).

We now illustrate the main motivation behind our investigation.

Let $n, m \ge 2$ be two integers. We can imagine a game that starts with the referee asking Alice and Bob to name two local GPTs A, B of fixed dimensions n, m, which they will have to manipulate later. Since the referee has no control over the experimental capabilities of the other two players, these are free to pick any A and B, subjected to the constraints dim $V_A = n$ and dim $V_B = m$. Once this choice has been made, it is communicated to the referee, who *physically constructs* A and B, combines them in a bipartite system AB that is a legitimate GPT but is elsewhere of their choice, selects an appropriate XOR game G over AB, and plays it with Alice and Bob. The goal of the referee is to make the global/local bias ratio of G as large as possible – ideally, one would aim for a global bias close to 1 and a local bias close to 0 – by suitably choosing the composite AB and the game G. Conversely, Alice and Bob wish to keep the bias ratio as close as possible to 1, by making the appropriate choice of local systems A and B. In other words, they want to be able to win the game with a reasonable probability without having to resort to global strategies. Given $n, m \ge 2$, how large is the global/local bias ratio the referee can hope to achieve?

Having specified the setting, we now proceed to perform the quantitative analysis that will ultimately lead us to identify our main object of study. Let us start by assuming that Alice and Bob have already named GPTs A, B, that the referee has also selected a composite AB, and that the only choice that remains to be done is that of the game G.

¹ In fact, one can pick the assisting distribution to reproduce directly the answers a, b the players have to give. For a classical XOR game defined by questions x, y and correct answers $c_{xy} \in \{0, 1\}$, it suffices to define the assisting probability distribution by p(ab|xy) = 1/2 if $a \oplus b = c_{xy}$ and 0 otherwise.

Start by observing that (3) implies that every element $z \in V_A \otimes V_B$ such that $||z||_{AB} \leq 1$ for some legitimate composite AB is such that $z = z_G$ for some game G on AB (with just two questions). Hence, we can equivalently parametrise XOR games with vectors z in the unit ball of $|| \cdot ||_{AB}$. The maximal global/local bias ratio the referee can hope to achieve then reads

$$\sup_{G} \frac{\beta_{ALL}(G)}{\beta_{LO}(G)} = \sup_{0 \neq z \in V_A \otimes V_B} \frac{\|z\|_{AB}}{\|z\|_{V_A \otimes_{\varepsilon} V_B}},$$
(25)

where the supremum on the l.h.s. is over all games on a fixed composite AB. Since the referee is also free to choose the optimal composite AB for given local systems A and B, we can also optimise over all composition rules. This is easily done by means of (12), and yields

$$\sup_{AB,G} \frac{\beta_{ALL}(G)}{\beta_{LO}(G)} = \sup_{0 \neq z \in V_A \otimes V_B} \frac{\|z\|_{V_A \otimes_\pi V_B}}{\|z\|_{V_A \otimes_\varepsilon V_B}} = \rho\left(V_A, V_B\right), \tag{26}$$

where the last step is an application of the definition (15). The above equation (26) is important because it connects the theory of XOR games over GPTs to that of tensor norms, and it can be used as a starting point to investigate some intrinsic aspects of the behaviour of information processing in bipartite systems. For instance, in [32] the optimal performance of data hiding against 'locally constrained sets of measurements' is connected to the quantity $\sup_{A,B} \rho(V_A, V_B)$, the supremum running over all GPTs A and B of fixed dimensions.² In the setting we study here, instead, Alice and Bob's goal is to minimise the global/local bias ratio in (26) by making a clever choice of A and B in the first place. Hence, the relevant quantity is

$$r_{\mathrm{bn}}(n,m) := \inf_{\substack{\dim A = n \\ \dim B = m}} \rho\left(V_A, V_B\right),\tag{27}$$

the infimum running over all GPTs A, B (equivalently, over all base norm spaces V_A , V_B) of fixed dimensions. If we find that $r_{bn}(n, m) > 1$ for all n, m, then there is a point in claiming that global strategies for XOR games perform better than local ones, independently of the underlying physical theories. If we manage to determine the asymptotic scaling of the quantity (27), we will even be able to make this statement quantitative. Comparing (27) with (16), it is elementary to observe that

$$r_{\rm bn}(n,m) \geqslant r(n,m) \tag{28}$$

for all $n, m \ge 2$, as the infimum that defines the r.h.s. is over all pairs of Banach spaces, while that on the l.h.s. includes only base norm spaces. In spite of this, thanks to the result of "Appendix B" we know that the two sides of (28) are in fact comparable. Namely, Lemma B.2 tells us that for all $n, m \ge 2$ one has

$$r_{\rm bn}(n,m) \leqslant 4 r(n,m) \,, \tag{29}$$

$$r_{\rm bn}(n,m) \leqslant 2 + r(n-1,m-1)$$
. (30)

In light of the above equivalences, in the rest of the paper we shall study the function r instead of r_{bn} . This simplifies our investigation considerably.

Before we present our results, let us comment on the optimisation over the composition rules performed in (26)–(27). While we argued above that ours may be most natural

² For convenience, here we are extending the definition of 'locally constrained sets of measurements' with respect to that given in [32], including also the scenario corresponding to an XOR game.

choice, it is conceivable to consider a modified scenario where Alice and Bob fix not only the systems *A*, *B*, but also the composite *AB*. Instead of (26), one should rather compute $\inf_{AB} \sup_{G} \beta_{ALL}(G)/\beta_{LO}(G)$, and then take also the infimum over *A* and *B* of fixed dimensions, as in (27). We will not consider further this alternative scenario. Remarkably, the choice of the composite is irrelevant when either *A* or *B* are classical theories. In this case, we are also able to give sharp estimates of the r.h.s. of (26) (see Remarks after Proposition 15 and Lemma 19).

2. Main Results

Throughout this section, we present our main results on the universal functions r(n, m)and $r_s(n)$ introduced in (16) and (17), respectively. The discussion of the elementary properties of these objects, as conducted in Sect. 1.2, left many natural questions open. For instance, *is it the case that* $r(n, m) \ge c > 1$ *for all* $n, m \ge 2$ *and for some universal constant* c? In other words, is it true that for any pair of Banach spaces X, Y of dimension at least 2 there is a tensor $z \in X \otimes Y$ for which $||z||_{X \otimes_{\pi} Y} \ge c||z||_{X \otimes_{\varepsilon} Y}$? Looking only at large enough dimensions, *how do* r(n, m) *and* $r_s(n)$ *behave asymptotically in* n, m? Some insight into this latter question was provided by Pisier [48], who showed that $r(n, m) \to \infty$ when min $\{n, m\} \to \infty$, but with no asymptotic growth explicitly stated. Our first result answers the first of the above questions in the affirmative.

Theorem 2. For any pair of Banach spaces X, Y with dim $X \ge 2$ and dim $Y \ge 2$, we

Theorem 2. For any pair of Banach spaces X, Y with dim $X \ge 2$ and dim $Y \ge 2$, we have $\rho(X, Y) \ge 19/18$. Equivalently, there exists a nonzero tensor $z \in X \otimes Y$ such that

$$\|z\|_{X\otimes_{\pi}Y} \geq \frac{19}{18} \|z\|_{X\otimes_{\varepsilon}Y}.$$

Consequently, the function r(n, m) defined in (16) satisfies

$$r(n,m) \ge \frac{19}{18} \quad \forall n,m \ge 2.$$
(31)

Theorem 2 also applies to infinite-dimensional spaces and shows that the injective and projective tensor products cannot be isometric. Our proof of Theorem 2 requires a variation on Auerbach's lemma that is susceptible to an intuitive geometrical interpretation (Lemma 16).

The following open problem asks whether the injective and projective tensor norms are always $\sqrt{2}$ apart. The value of $\sqrt{2}$ would be optimal, since $\rho(\ell_1^2, \ell_2^2) = \sqrt{2}$ (for a proof of this fact, see (59)).

Problem 3. Is it true that for any pair of Banach spaces X, Y with dim X, dim $Y \ge 2$, we have $\rho(X, Y) \ge \sqrt{2}$? In other words, does there always exist a nonzero tensor $z \in X \otimes Y$ such that

$$\|z\|_{X\otimes_{\pi}Y} \ge \sqrt{2} \|z\|_{X\otimes_{\varepsilon}Y}$$

$$(32)$$

We now move on to the analysis of the asymptotic behaviour of the function $r_s(n)$. Developing functional analysis techniques from [49], we arrive at a lower estimate of $\rho(X, X)$ involving the Banach–Mazur distance of X from the Hilbert space of the same dimension (Theorem 17). Since by John's theorem (Theorem C.7) this cannot exceed the square root of the dimension, we deduce the following estimate, which is sharp up to logarithmic factors (compare with (20)). **Theorem 4.** The function r_s defined in (17) satisfies

$$\sqrt{2n} \ge r_s(n) \ge c \, \frac{\sqrt{n}}{(\log n)^3} \tag{33}$$

for some universal constant c > 0.

Note. Throughout the paper we will denote universal (always strictly positive) constants by c, C, C' etc. Unless explicitly indicated, these symbols do not necessarily refer to the same numerical values when they appear in different formulae.

The investigation of the function r(n, m) poses more substantial technical hurdles. For simplicity, we start by looking at r(n, n). Our main result is the following.

Theorem 5. *The function r defined in* (16) *satisfies*

$$\sqrt{n} \ge r(n,n) \ge c \, \frac{n^{1/6}}{(\log n)^{4/3}} \tag{34}$$

for some universal constant c > 0.

The above estimate shows in particular that r(n, n) grows at least as a power of n. In turn, this implies that *global strategies for XOR games are intrinsically asymptotically (much) more efficient than local ones*, which was one of our main claims. Our proof of Theorem 5 rests upon two main ingredients: (1) a lower estimate for $\rho(\ell_1^n, X)$, $\rho(\ell_{\infty}^n, X)$ and $\rho(\ell_2^n, X)$ when dim $X \ge n$ (Lemma 19), which can be handled using known facts about *p*-summing norms; and (2) a 'trichotomy theorem' inspired by previous results in [7,50], which states that every Banach space hosts sufficiently well-behaved subspaces on which the norm is similar enough to either an ℓ_1 -norm, or an ℓ_{∞} -norm, or a Euclidean norm (Theorem 20).

Due to the technical complexity of managing many different estimates simultaneously, for the case $n \neq m$ we could not obtain an exponent as good as 1/6. However, we were nevertheless able to ensure that there is power law scaling in min $\{n, m\}$.

Theorem 6. For all $n, m \ge 2$, the function r(n, m) satisfies

$$\sqrt{\min\{n,m\}} \ge r(n,m) \ge c \, \frac{\min\{n,m\}^{1/8}}{\log\min\{n,m\}} \tag{35}$$

for some universal constant c > 0.

The exponents we obtained in Theorems 5 and 6 are unlikely to optimal. We present the following problem concerning the scaling of the projective/injective ratio.

Problem 7. Does there exist a universal constant c > 0 such that

$$r(n,m) \ge c \min\{n,m\}^{1/2} \tag{36}$$

for all positive integers n, m? In other words, for all pairs of finite-dimensional Banach spaces X, Y, does there exist a nonzero tensor $z \in X \otimes Y$ such that

$$||z||_{X\otimes_{\pi}Y} \ge c \min\{\dim X, \dim Y\}^{1/2} ||z||_{X\otimes_{\varepsilon}Y} ?$$
(37)

While the value of r(n, m) grows with the dimension, new phenomena appear when considering infinite-dimensional spaces. Indeed, a famous construction by Pisier [8], solving negatively a conjecture by Grothendieck, entails that there exists an infinite-dimensional Banach space X such that $\rho(X, X) < \infty$. Using the information from Lemma B.1, we conclude that the same behaviour occurs in the realm of GPTs: there exist infinite-dimensional GPTs A, B such that local and global strategies are equivalent up to a universal constant in any possible composite AB.

The study of asymptotic behaviours in the general setting should not distract us from the fact that certain GPT models are of prime importance because of their compliance with known physics. Therefore, one of our results is the determination of the quantity $\rho(X, Y)$ when X, Y are the base norm spaces corresponding to quantum mechanical systems, i.e. the Banach spaces $S_1^{k,sa}$. The following constitutes a notable improvement over [51, Lemma 20], [52, Theorem 15] and [32, Eq. (72)], as detailed below.

Theorem 8. Denoting by $S_1^{k,sa}$ the space of $k \times k$ Hermitian matrices endowed with the trace norm, the best constant of domination of $\|\cdot\|_{X\otimes_{\varepsilon}Y}$ over $\|\cdot\|_{X\otimes_{\pi}Y}$ on $S_1^{n,sa} \otimes S_1^{m,sa}$ satisfies

$$c \min\{n, m\}^{3/2} \le \rho\left(S_1^{n, \mathrm{sa}}, S_1^{m, \mathrm{sa}}\right) \le C \min\{n, m\}^{3/2}$$
 (38)

for some constants C, c > 0. More precisely, we have the following estimate for the upper bound in the above relation:

$$\rho\left(S_1^{n,\text{sa}}, S_1^{m,\text{sa}}\right) \leqslant 4\min\{n, m\}^{3/2} - 2\sqrt{2}(\sqrt{2} - 1)\sqrt{\min\{n, m\}}.$$
(39)

Note. When comparing (38) with the other estimates on r(n, m) that we presented throughout this section, one should remember that the dimension of the space $S_1^{n,sa}$ is n^2 rather than n. Hence, curiously, for $X = S_1^{n,sa}$ the quantity $\rho(X, X) \approx n^{3/2}$ is of the same order as the geometric mean between the theoretical minimum $r_s(n^2) \approx n$ and the absolute maximum $R(n^2, n^2) = n^2$.

Although we will not consider complex spaces in this work, it is worth mentioning that the same estimate holds (with a slight modification in the constants) if the real space $S_1^{k,sa}$ is replaced by the complex space S_1^k of $k \times k$ matrices endowed with the trace norm.

Among other things, the above Theorem 8 enables us to give a new upper bound on the maximal efficiency of quantum mechanical data hiding against local measurements, quantitatively encoded in the data hiding ratio function $R_{LO}(n, m)$ defined in [32].

Corollary 9. For all Hermitian matrices Z acting on a bipartite system $\mathbb{C}^n \otimes \mathbb{C}^m$, we have

$$\|Z\|_{S_{1}^{n,\mathrm{sa}}\otimes_{\varepsilon}S_{1}^{m,\mathrm{sa}}} \leqslant \|Z\|_{\mathrm{LO}} \leqslant \|Z\|_{1} \leqslant \|Z\|_{S_{1}^{n,\mathrm{sa}}\otimes_{\pi}S_{1}^{m,\mathrm{sa}}} \leqslant 4\min\{n,m\}^{3/2}\|Z\|_{S_{1}^{n,\mathrm{sa}}\otimes_{\varepsilon}S_{1}^{m,\mathrm{sa}}},$$
(40)

where $\|\cdot\|_{LO}$ is the distinguishability norm under local measurements [52]. In other words, the data hiding ratio against local measurements can be upper bounded as

$$R_{\rm LO}(n,m) \leqslant 4\min\{n,m\}^{3/2}$$
. (41)

The above result improves upon several previously known estimates. In [51, Lemma 20], an inequality analogous to (41) was shown, that featured an exponent 2 on the r.h.s.; moreover, the relation proven in [52, Theorem 15] implies that $R_{\text{LO}}(n, m) \leq \sqrt{153} nm$, which can be worse than (41) e.g. when $n^2 \leq m$ or $m^2 \leq n$.

3. First Bounds on the π/ϵ Ratio

3.1. Some notions of functional analysis. We start by reminding the reader of some facts in elementary linear algebra. Given a pair of finite-dimensional vector spaces X, Y, there is a canonical isomorphism between the tensor product space $X \otimes Y$ and the space of linear maps $X^* \to Y$. We will write this correspondence as

$$X \otimes Y \ni z \longmapsto \widetilde{z} \in \mathcal{L}(X^*, Y) .$$
⁽⁴²⁾

Note that one has $F(z) = \tilde{z}^*$, where on the l.h.s. we have the flip operator $F : X \otimes Y \to Y \otimes X$ defined by $F(x \otimes y) := y \otimes x$, while on the r.h.s. $(\cdot)^*$ stands for the adjoint (transposition) operation $\mathcal{L}(X^*, Y) \to \mathcal{L}(Y^*, X)$. It is also easy to see that one has

 $w(z) = \operatorname{tr}\left[\widetilde{z}^*\widetilde{w}\right] = \operatorname{tr}\left[\widetilde{z}^{}\widetilde{w}^*\right] \quad \forall z \in X \otimes Y, \quad \forall w \in (X \otimes Y)^* = X^* \otimes Y^*, \quad (43)$

where tr denotes the trace.

In this paper we are interested in tensor products of finite-dimensional real Banach spaces, so from now on X and Y will denote a pair of such objects. We already encountered the concepts of injective and projective tensor products [see (10) and (11)]. Below we discuss some elementary properties of these constructions. For a start, injective and projective norm are dual to each other, in the sense that

$$\|\cdot\|_{(X\otimes_{\varepsilon}Y)^{*}} = \|\cdot\|_{X^{*}\otimes_{\pi}Y^{*}}.$$
(44)

By means of the correspondence (42), it is possible to translate tensor norms into the language of operators. One has

$$\|z\|_{X\otimes_{\varepsilon}Y} = \|\widetilde{z}: X^* \to Y\|, \qquad (45)$$

$$\|z\|_{X\otimes_{\pi}Y} = \|\widetilde{z} : X^* \to Y\|_N, \qquad (46)$$

where $\|\cdot\|_N$ is the so-called **nuclear norm** [53, Section 8], which may be more familiar to some readers. This is particularly transparent for the tensor product of two Euclidean spaces: if $X = \ell_2^m$, $Y = \ell_2^n$, then $\|z\|_{X \otimes_{\varepsilon} Y} = \|\widetilde{z}\|_{\infty}$ and $\|z\|_{X \otimes_{\pi} Y} = \|\widetilde{z}\|_1$, where $\|M\|_{\infty}$ and $\|M\|_1$ denote the operator norm and the trace norm of a matrix M. The following well-known lemma covers special cases which are simple: the projective tensor product with ℓ_1^n and the injective tensor product with ℓ_{∞}^n .

Lemma 10. If $(e_i)_{1 \le i \le n}$ denotes the canonical basis of \mathbb{R}^n , then for every vectors x_1, \dots, x_n in a Banach space X,

$$\left\|\sum_{i=1}^{n} e_{i} \otimes x_{i}\right\|_{\ell_{1}^{n} \otimes_{\pi} X} = \sum_{i=1}^{n} \|x_{i}\|,$$
(47)

$$\left\|\sum_{i=1}^{n} e_i \otimes x_i\right\|_{\ell_{\infty}^n \otimes_{\varepsilon} X} = \max_{1 \leqslant i \leqslant n} \|x_i\|.$$
(48)

Proof. The upper bound in (47) is immediate from the definition. Conversely, given a decomposition of $\sum e_i \otimes x_i$ as $\sum \alpha^k \otimes \xi^k$, we have

$$\sum_{k} \|\alpha^{k}\|_{1} \|\xi^{k}\| = \sum_{k} \sum_{i=1}^{n} |\alpha_{i}^{k}| \|\xi^{k}\| \ge \sum_{i=1}^{n} \left\|\sum_{k} \alpha_{i}^{k} \xi^{k}\right\| = \sum_{i=1}^{n} \|x_{i}\|$$

and equality in (47) follows once we take the infimum over decompositions. The proof of (48) is immediate once we realize that the supremum in the definition of the injective norm can be restricted to extreme points of the unit ball. $\hfill \Box$

The **Banach–Mazur distance** between two normed spaces X, Y with the same finite dimension is defined as [53]

$$d(X, Y) := \inf_{u} \left\{ \|u : X \to Y\| \cdot \|u^{-1} : Y \to X\| \right\},$$
(49)

where the infimum is taken over invertible linear maps u from X to Y. It satisfies the multiplicative version of the triangle inequality, i.e. $d(X, Z) \leq d(X, Y) d(Y, Z)$. Another elementary property is the fact that $d(X, Y) = d(X^*, Y^*)$. When $d(X, Y) \leq \lambda$, we say that X is λ -isomorphic to Y. Similarly, X is λ -Euclidean if it is λ -isomorphic to $\ell_2^{\dim X}$.

The Banach–Mazur distance only makes sense for a pair of spaces of equal dimension. When *X*, *Y* are normed spaces such that dim $X \leq \dim Y$, one may define as a substitute the **factorisation constant** of *X* through *Y* as

$$f(X, Y) := \inf_{u, v} \{ \|u : X \to Y\| \cdot \|v : Y \to X\| : vu = Id_X \}.$$
(50)

We point that d(X, Y) = f(X, Y) if dim $X = \dim Y$. Moreover, f(X, Y) is finite if and only if dim $X \leq \dim Y$ (otherwise the above infimum is over an empty set). In order to circumvent this restriction, it is sometimes relevant to consider a relaxed version of the above quantity, where we allow the factorisation to be realised only after averaging. This leads to the definition of the **weak factorisation constant** as

wf(X, Y) :=
$$\inf_{u,v} \{ \mathbf{E} [\| u : X \to Y \| \cdot \| v : Y \to X \|] : \mathbf{E} [vu] = \mathrm{Id}_X \},$$
 (51)

where now the infimum is taken over pairs of operator-valued random variables (u, v). We get the inequality wf $(X, Y) \leq f(X, Y)$ by restricting the infimum to constant random variables. Note also that the quantity wf(X, Y) is well-defined without any restriction on the dimensions of *X* and *Y*. It is easy to verify that the factorisation constants dualise, in the sense that

$$f(X, Y) = f(X^*, Y^*),$$
 (52)

$$wf(X, Y) = wf(X^*, Y^*).$$
 (53)

We may also consider a symmetric variant of the weak factorisation constant, called the **weak Banach–Mazur distance** and defined as

$$wd(X, Y) = \max\{wf(X, Y), wf(Y, X)\}.$$

The family of all equivalence classes of *n*-dimensional normed spaces up to isometries can be turned into a compact metric space by the introduction of the distance log d. We now review some classical facts about the geometry of such space, called the **Banach–Mazur compactum** of dimension *n*. For a more complete introduction, we refer the reader to the excellent monograph [53]. A general upper bound valid for any *n*-dimensional normed space is the estimate following from John's theorem (Theorem C.7)

$$d(X, \ell_2^n) \leqslant \sqrt{n}. \tag{54}$$

As a consequence of this and the multiplicative triangle inequality, for any pair of n-dimensional spaces X, Y, we have

$$d(X,Y) \leqslant n. \tag{55}$$

This bound is essentially sharp: Gluskin proved, via a random construction, the existence of spaces *X*, *Y* such that $d(X, Y) \ge cn$ for some c > 0 [54]. However, the estimate (55) can be improved in many specific cases. In particular, a question that is of relevance to us is that of the distance between a space and its dual: it was proved by Bourgain and Milman [49] that whenever dim(*X*) = *n*, we have

$$d(X, X^*) \leqslant C n^{5/6} \log^C n.$$
(56)

In a similar vein, (55) can be improved if we switch to the weak Banach–Mazur distance: a result by Rudelson [7] asserts that

$$wd(X,Y) \leqslant Cn^{13/14} \log^C n \tag{57}$$

whenever $\dim(X) = \dim(Y) = n$. We also point out that the exact diameter of the Banach–Mazur compactum is known in dimension 2: namely,

$$\max_{\substack{\dim X=2\\\dim Y=2}} d(X,Y) = \frac{3}{2},$$
(58)

and the equality is achieved iff the unit balls of X and Y are the images of a square and a regular hexagon through a linear (invertible) map [55].

3.2. Basic properties. We start the investigation of the elementary properties of the parameter $\rho(X, Y)$ by presenting a simple calculation in the case $X = \ell_1^n$, $Y = \ell_2^m$, where $n \leq m$ (we point that the calculation can be rephrased in terms of 1-summing norms, see Proposition 13). For a matrix $z \in \ell_1^n \otimes \ell_2^m$, we write

$$\begin{aligned} \|z\|_{\ell_{1}^{n}\otimes_{\varepsilon}\ell_{2}^{m}} &= \left(\sup_{\sigma\in\{\pm1\}^{n}}\sum_{j=1}^{m}\left(\sum_{i=1}^{n}z_{ij}\sigma_{i}\right)^{2}\right)^{1/2} \\ &\geqslant \left(\mathbf{E}_{\sigma}\sum_{j=1}^{m}\left(\sum_{i=1}^{n}z_{ij}\sigma_{i}\right)^{2}\right)^{1/2} \\ &= \left(\sum_{j=1}^{m}\sum_{i=1}^{n}z_{ij}^{2}\right)^{1/2} \\ &\geqslant \frac{1}{\sqrt{n}}\sum_{i=1}^{n}\left(\sum_{j=1}^{m}z_{ij}^{2}\right)^{1/2} \\ &= \frac{1}{\sqrt{n}}\|z\|_{\ell_{1}^{n}\otimes_{\pi}\ell_{2}^{m}}, \end{aligned}$$

where to obtain the first inequality we randomised over σ , assuming that $\sigma_1, \ldots, \sigma_n$ are i.i.d. ± 1 Bernoulli random variables, and the last equality uses Lemma 10. The above computation shows that $||z||_{\ell_1^n \otimes_{\pi} \ell_2^m} \leq \sqrt{n} ||z||_{\ell_1^n \otimes_{\varepsilon} \ell_2^m}$, implying that $\rho(\ell_1^n, \ell_2^m) \leq \sqrt{n}$. That this upper bound is in fact tight can be seen by considering the matrix *z* with entries $z_{ij} = \delta_{i,j}$ (i.e. the identity if m = n), for which $||z||_{\ell_1^n \otimes_\pi \ell_2^m} = n$ (from Lemma 10) and $||z||_{\ell_1^n \otimes_\pi \ell_2^m} = \sqrt{n}$ (as seen from the above computation). We conclude that

$$\rho(\ell_1^n, \ell_2^m) = \sqrt{n} \quad \forall n \leqslant m \,, \tag{59}$$

entailing the upper bound (19) on r(n, m).

We now move on to investigating some more general properties of the function $\rho(X, Y)$. We start by stating a very useful reformulation in terms of operators rather than tensors.

Lemma 11. For any pair of finite-dimensional normed spaces X, Y, we have

$$\rho(X, Y) = \sup \left\{ \operatorname{tr}(vu) : \|u : X \to Y^*\| \leq 1, \|v : Y^* \to X\| \leq 1 \right\}.$$
(60)

Proof. The statement follows from the properties of injective and projective norms under the correspondence (42). We have

$$\begin{split} \rho(X,Y) &= \sup_{0 \neq z \in X \otimes Y} \frac{\|z\|_{X \otimes_{\pi} Y}}{\|z\|_{X \otimes_{\varepsilon} Y}} \\ &= \sup_{0 \neq z \in X \otimes Y} \sup_{0 \neq w \in X^* \otimes Y^*} \frac{w(z)}{\|w\|_{X^* \otimes_{\varepsilon} Y^*} \|z\|_{X \otimes_{\varepsilon} Y}} \\ &= \sup \left\{ w(z) : \|w\|_{X^* \otimes_{\varepsilon} Y^*} \leqslant 1, \|z\|_{X \otimes_{\varepsilon} Y} \leqslant 1 \right\} \\ &= \sup \left\{ tr\left[(\widetilde{z})^* \widetilde{w} \right] : \|\widetilde{w} : X \to Y^*\| \leqslant 1, \|\widetilde{z} : X^* \to Y\| \leqslant 1 \right\} \\ &= \sup \left\{ tr\left[vu \right] : \|u : X \to Y^*\| \leqslant 1, \|v : Y^* \to X\| \leqslant 1 \right\}. \end{split}$$

The above passages can be justified as follows: 1: we used the duality relation (44); 2: we applied (43), (45) and (46); 3: we just renamed $v := (\tilde{z})^*$ and $u := \tilde{w}$. \Box

The following proposition gathers several estimates of the function $\rho(X, Y)$.

Proposition 12. Let X, X' and Y be finite-dimensional normed spaces. Then

$$\rho(X, Y) = \rho(Y, X) = \rho(X^*, Y^*) = \rho(Y^*, X^*),$$
(61)

$$\rho(X, X^*) = \dim(X), \qquad (62)$$

$$\rho(X',Y) \leqslant \operatorname{wf}(X',X)\rho(X,Y) \tag{63}$$

$$\leqslant f(X', X)\rho(X, Y), \tag{64}$$

$$\rho(X', Y) \leqslant d(X', X)\rho(X, Y) \quad (assuming \ \dim(X) = \dim(X')), \tag{65}$$

$$\rho(X,Y) \geqslant \frac{\dim X}{\operatorname{wf}(X,Y^*)},\tag{66}$$

$$\rho(X, Y) \leqslant \min\left\{\dim(X), \dim(Y)\right\}.$$
(67)

In particular, when $\dim(X) = \dim(Y) = n$ we have

$$\rho(X,Y) \ge \frac{n}{\operatorname{wd}(X,Y^*)} \ge \frac{n}{\operatorname{d}(X,Y^*)} \,. \tag{68}$$

Proof. The identities (61) can be obtained for instance from (60) by exchanging the role of u and v and/or taking their duals u^* , v^* . Remember that one has $tr(uv) = tr(vu) = tr(u^*v^*) = tr(v^*u^*)$. Also (62) is elementary, and follows by taking $u = v = Id_X$ in (60).

To show (63), pick some operators $u : X' \to Y^*$ and $v : Y^* \to X'$ of norm no larger than 1, and consider random variables $u' : X' \to X$ and $v' : X \to X'$ such that $\mathbf{E}[v'u'] = \mathrm{Id}_{X'}$. Since for all realisations of u' and v' the operator $\frac{uv'}{\|v'\|} : X \to Y^*$ has norm no larger than 1, and an analogous reasoning holds for $\frac{u'v}{\|u'\|} : Y^* \to X$, using (60) we deduce that

$$\frac{\operatorname{tr}(u'vuv')}{\|u'\|\|v'\|} \leqslant \rho(X,Y) \,.$$

Then we can write

$$tr(vu) = tr (vu \mathbf{E}[v'u'])$$

= $\mathbf{E} [tr (vuv'u')]$
= $\mathbf{E} [tr (u'vuv')]$
= $\mathbf{E} [||u'||||v'|| \frac{tr(u'vuv')}{||u'||||v'||}]$
 $\leq \rho(X, Y) \mathbf{E} [||u'||||v'||].$

Taking the supremum over u, v and the infimum over u', v' subjected to the above constraints, and using (51) and (60), we finally obtain (63). Since wf $(X', X) \leq f(X', X)$, (64) follows as well. To prove (65), note that d(X, X') = f(X, X') whenever $\dim(X) = \dim(X')$. Note that (66) also follows immediately:

$$\dim X \stackrel{1}{=} \rho(X, X^*) \tag{69}$$

$$\stackrel{2}{\leqslant} \operatorname{wf}(X, Y^*)\rho(Y^*, X^*) \tag{70}$$

$$\stackrel{3}{=} \operatorname{wf}(X, Y^*)\rho(X, Y) \tag{71}$$

where: 1: follows from (62) and (61); 2: is an application of (63); and 3: is again a consequence of (61).

The estimate (67) was proved in [32, Proposition 21] (compare with (18)) as a consequence of Auerbach's lemma [56, Vol. I, Sec. 1.c.3]. An alternative proof can be given using the following fact from linear algebra (left as an exercise for the reader): every linear map w on \mathbb{R}^n that is a contraction for some norm satisfies tr(w) $\leq \operatorname{rank}(w)$. To recover (67), apply this observation to the composite map w = vu, with u, v as in (60). \Box

Remark. It follows in particular from (64) that any upper bound on $\rho(X, Y)$ is also valid for 1-complemented subspaces of *X* and *Y*. Recall that a subspace $X' \subseteq X$ is λ -complemented if there is a surjective projection $P : X \to X'$ with $||P|| \leq \lambda$. The complementation hypothesis cannot be omitted. To give a concrete example, consider a *cn*-dimensional subspace $X \subseteq \ell_1^n$ with $d(X, \ell_2^{cn}) \leq C$ (existence of such a subspace is well known and follows for example from Dvoretzky–Milman theorem (Theorem C.6) since $k_*(\ell_1^n) \geq cn$): we have $\rho(X, X) = \Theta(n)$ while $\rho(\ell_1^n, \ell_1^n) = \Theta(\sqrt{n})$.

Remark. From (65) we see in particular that the function $\rho(\cdot, \cdot)$ defined on the product of the Banach–Mazur compacta of dimensions *n* and *m* is continuous (with respect to the product metric). In particular, this implies that the infima in (16) and (17) are always achieved. We will make use of this fact without further mention in what follows.

We point out that weaker versions of Theorems 4 and 5 follow easily by combining Proposition 12 and 'off-the-shelf' results. More precisely, the lower bound $r_s(n) \ge cn^{1/6}/(\log n)^C$ is an immediate consequence of (68) and (56), and the lower bound $r(n, n) \ge cn^{1/14}/(\log n)^C$ follows from (66) and (57). Interestingly, the special case n = m of Problem 7 would follow from (68) if one could prove Rudelson's conjecture that wf(*X*, *Y*) $\le C\sqrt{n}$ for all *n*-dimensional Banach spaces *X*, *Y*.

In the case where one of the spaces is ℓ_1^n , the quantity $\rho(X, Y)$ can be rephrased in terms of 1-summing norms (the quantities $\pi_1^{(n)}(u)$ and $\pi_1(u)$ are defined in "Appendix 5").

Proposition 13. For every finite-dimensional normed space X, we have

$$\rho(\ell_1^n, X) = \pi_1^{(n)}(\mathrm{Id}_X) \leqslant \pi_1(\mathrm{Id}_X).$$

Proof. Using Lemma 11, we have

$$\rho(\ell_1^n, X) = \rho(X, \ell_1^n) = \sup\{\operatorname{tr}(vu) : \|u : X \to \ell_\infty^n\| \le 1, \|v : \ell_\infty^n \to X\| \le 1\}.$$
(72)

We rewrite the norms which appear in (72) in a more tangible way. If $v : \ell_{\infty}^n \to X$ and $v_i = v(e_i)$, then (72)

$$\|v\| = \sup\left\{\left\|\sum_{i=1}^{n} \varepsilon_i v_i\right\| : \varepsilon_i = \pm 1\right\}.$$

Next, if $u : X \to \ell_{\infty}^{n}$ is given by $x \to (\langle f_{i}, x \rangle)_{i=1}^{n}$ for some $f_{i} \in X^{*}$, then $||u|| = \max_{i} ||f_{i}||$. Finally, if u and v are as above, then $\operatorname{tr}(vu) = \operatorname{tr}(uv) = \sum_{i=1}^{n} \langle f_{i}, v_{i} \rangle$. Combining these we are led to

$$\rho(\ell_1^n, X) = \sup\left\{\sum_{i=1}^n \|v_i\| : v_i \in X, \max_{\varepsilon_i = \pm 1} \left\|\sum_{i=1}^n \varepsilon_i v_i\right\| \le 1\right\}.$$
 (73)

Comparing with (C4) and using the relation

$$\max_{\varepsilon_i=\pm 1} \left\| \sum_{i=1}^n \varepsilon_i v_i \right\| = \sup_{\phi \in B_{X^*}} \sum_{i=1}^n |\phi(v_i)|$$

shows that $\rho(\ell_1^n, X) = \pi_1^{(n)}(\mathrm{Id}_X)$ (the general inequality $\pi_1^{(n)}(\cdot) \leq \pi_1(\cdot)$ is immediate from the definitions). \Box

With this connection at hand, we are able to give an improved upper bound on the parameter $r_s(n)$.

Proposition 14. For every $n \ge 1$, we have

$$r_s(n) \leqslant \rho(\ell_1^n, \ell_1^n) \leqslant \pi_1(\mathrm{Id}_{\ell_1^n}) = \frac{n}{\mathbf{E}\left|\sum_{i=1}^n \varepsilon_i\right|} \overset{n \to \infty}{\sim} \sqrt{\frac{\pi n}{2}}$$

where $\varepsilon_1, \ldots, \varepsilon_n \in \{\pm 1\}$ are independent random variables with $\mathbf{P}(\varepsilon_i = 1) = \mathbf{P}(\varepsilon_i = -1) = 1/2$.

Proof. This is a simple consequence of Proposition 13 and (C5). \Box

3.3. Universal lower bounds. In this section we prove that the injective and projective tensor products of any two Banach spaces cannot be isometric, unless one of them is 1-dimensional. Going further, Problem 3 asks whether the injective and projective norms are always at least $\sqrt{2}$ apart. As a partial answer, we present a proof in the special case when one of the spaces is ℓ_{∞}^n (Proposition 15). Then, we solve a weaker version of Problem 3, with $\sqrt{2}$ replaced by the value 19/18 (Theorem 2). Finally, we discuss the special case when both dimensions are equal to 2.

Proposition 15. *If Y is a Banach space with* $dim(Y) \ge 2$ *, then for any* $n \ge 2$ *,*

$$\rho(\ell_{\infty}^n, Y) \geqslant \sqrt{2}.$$

Proof. Since ℓ_{∞}^2 is 1-complemented in ℓ_{∞}^n , in view of (64), it suffices to consider n = 2. In that case, there are explicit formulas for both the projective and injective norms: for any $z = e_1 \otimes y_1 + e_2 \otimes y_2 \in \ell_{\infty}^2 \otimes Y$, we have

$$\|z\|_{\ell^2_{\infty}\otimes_{\pi}Y} = \frac{1}{2} \left(\|y_1 + y_2\| + \|y_1 - y_2\| \right) \text{ and } \|z\|_{\ell^2_{\infty}\otimes_{\epsilon}Y} = \max\left\{ \|y_1\|, \|y_2\| \right\}.$$

Both formulas can be derived from Lemma 10, the first using the fact that the map $(a, b) \mapsto (a+b, a-b)$ is an isometry between ℓ_1^2 and ℓ_{∞}^2 . It remains to justify that any Banach space Y contains two vectors y_1 and y_2 such that

$$\|y_1 + y_2\| + \|y_1 - y_2\| \ge 2\sqrt{2} \max\{\|y_1\|, \|y_2\|\}.$$
(74)

This follows from properties of the so-called **modulus of uniform convexity** of *Y*, a real function defined for $\varepsilon \in [0, 2]$ by

$$\delta_Y(\varepsilon) = \inf \left\{ 1 - \frac{\|y_1 + y_2\|}{2} : \|y_1\| = \|y_2\| = 1, \|y_1 - y_2\| \ge \varepsilon \right\}.$$

It is known [57] that for any Banach space *Y* and any $\varepsilon \in [0, 2]$, we have $\delta_Y(\varepsilon) \leq 1 - \sqrt{1 - \varepsilon^2/4}$ (the value obtained for a Euclidean space). Applying this inequality with $\varepsilon = \sqrt{2}$ shows the existence of unit vectors y_1 , y_2 such that $||y_1 - y_2|| = \sqrt{2}$ and $||y_1 + y_2|| \ge \sqrt{2}$, and therefore (74) is satisfied. \Box

Remark. An immediate consequence of Proposition 15 is that there is always a gap at least as large as $\sqrt{2}$ between local and global bias for XOR games played over a system *AB* in which either *A* or *B* is a classical theory (7).

We now proceed to establish the universal lower bound $r(n, m) \ge 19/18$, formalised earlier as Theorem 2. Our main technical tool is the following variant of Auerbach's lemma, illustrated in Fig. 1.

Lemma 16. Let X be a Banach space of dimension at least 2. Then there exist vectors $e_1, e_2 \in X, e_1^*, e_2^* \in X^*$ such that for any $i, j \in \{1, 2\}$ we have $||e_i||_X = ||e_j^*||_{X^*} = 1$, $e_i^*(e_i) = \delta_{i,j}$, and moreover

$$||e_1 + e_2||_X \leq 3/2.$$

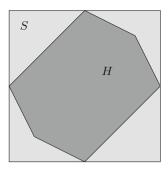


Fig. 1. Any centrally symmetric planar convex body has a linear image K satisfying $H \subseteq K \subseteq S$, where S is the square $[-1, 1]^2$ and H is the hexagon with vertices $(\pm 1, 0), (0, \pm 1)$ and $(\pm 2/3, \pm 2/3)$

Proof. It is enough to prove the lemma when dim X = 2, since the general case follows by considering any 2-dimensional subspace $Y \subseteq X$ and extending the linear forms.

Suppose now dim X = 2. Without loss of generality we may assume that $X = (\mathbf{R}^2, \|\cdot\|_X)$, and identify as well X^* with \mathbf{R}^2 . By applying a suitable linear transformation, we may assume that the variational problem max{ $|\det(f, g)| : f, g \in B_{X^*}$ } is achieved when $(f, g) = (e_1, e_2)$, the canonical basis. It is clear that $||e_1||_{X^*} = ||e_2||_{X^*} = 1$, and one checks that $||e_1||_X = ||e_2||_X = 1$. Let us show this explicitly for e_1 . On the one hand, $1 = \langle e_1, e_1 \rangle \leq ||e_1||_{X^*} ||e_1||_X = ||e_1||_X$. On the other hand, if $||e_1||_X > 1$ one could find a vector $f' \in B_{X^*}$ such that $\langle f', e_1 \rangle > 1$, which would imply that $|\det(f', e_2)| > 1 = |\det(e_1, e_2)|$, in contradiction with the assumption that the pair (e_1, e_2) achieves the maximum in the above variational expression.

Define $\alpha = ||e_1 + e_2||_X$ and $\beta = ||e_1 - e_2||_X$, and let $\phi, \psi \in B_{X^*}$ such that $\phi(e_1 + e_2) = \alpha$ and $\psi(e_1 - e_2) = \beta$. Write $\phi = (\phi_1, \phi_2)$ and $\psi = (\psi_1, \psi_2)$, so that $\alpha = \phi_1 + \phi_2$ and $\beta = \psi_1 - \psi_2$. We compute

$$1 \ge \det(\psi, \phi) = \phi_2 \psi_1 - \phi_1 \psi_2 \ge \alpha + \beta - 2.$$

To derive the last inequality, note that

$$\phi_2\psi_1 - \phi_1\psi_2 - (\alpha + \beta - 2) = (1 - \phi_2)(1 - \psi_1) + (1 - \phi_1)(1 + \psi_2)$$

is nonnegative since $|\phi_i| \leq 1$ and $|\psi_j| \leq 1$. We proved that $\alpha + \beta \leq 3$, and therefore either $\alpha \leq 3/2$ or $\beta \leq 3/2$. In the first case the conclusion is immediate; in the second case it suffices to replace e_2 by $-e_2$. \Box

Proof of Theorem 2. Apply Lemma 16 to both X and Y, and consider the tensor

$$z = 5e_1 \otimes e_1 + 5e_1 \otimes e_2 + 5e_2 \otimes e_1 - 4e_2 \otimes e_2 \in X \otimes Y.$$

Consider also

$$w^* = e_1^* \otimes e_1^* + e_1^* \otimes e_2^* + e_2^* \otimes e_1^* - e_2^* \otimes e_2^* \in X^* \otimes Y^*.$$

Since the linear forms e_i^* are bounded in absolute value by 1 on the unit ball, an argument following closely the proof of the CHSH inequality [44] shows that $||w^*||_{X^*\otimes_{\varepsilon}Y^*} \leq 2$. Together with the fact that $w^*(z) = 19$, this implies that

$$\|z\|_{X\otimes_{\pi}Y} \geqslant \frac{19}{2}.$$
(75)

It remains to upper bound $||z||_{X \otimes_{\varepsilon} Y}$. Given $\phi \in B_{X^*}$ and $\psi \in B_{Y^*}$, consider the numbers

$$a = \phi(e_1), \ b = \phi(e_2), \ c = \psi(e_1), \ d = \psi(e_2).$$

Both pairs (a, b) and (c, d) belong to the hexagon

$$H = \left\{ (x, y) \in \mathbf{R}^2 : |x| \leq 1, |y| \leq 1, |x + y| \leq 3/2 \right\}.$$

Under these constraints it can be proved that

$$5ac + 5ad + 5bc - 4bd \leq 9$$

Indeed, it suffices to verify this inequality when (a, b) and (c, d) are extreme points of *H*; this yields a total of 36 different combinations to check. Finally, we have

$$\|z\|_{X\otimes_{\varepsilon}Y} = \sup_{\phi\in B_{X^*}, \,\psi\in B_{Y^*}} (\phi\otimes\psi)(z)\leqslant 9.$$
(76)

Combining (75) with (76) gives $||z||_{X\otimes_{\pi}Y} \ge \frac{19}{18} ||z||_{X\otimes_{\varepsilon}Y}$, as needed. \Box

Remark. The proof of Lemma 16 gives more information, namely that

$$\|e_1 + e_2\|_X \leqslant \alpha, \ \|e_1 - e_2\|_X \leqslant \beta$$

for some real numbers α , $\beta \in [1, 2]$ such that $\alpha + \beta \leq 3$. This extra information can presumably be used to improve the lower bound in Theorem 2 to $||z||_{X\otimes_{\pi}Y} \geq \frac{8}{7}||z||_{X\otimes_{\varepsilon}Y}$ for an appropriate choice of *z* depending on α , β . However, since our arguments for that would rely heavily on computer assistance (and since the bound 8/7 is unlikely to be optimal), we do not present them.

Before we move on, let us discuss the special case of 2-dimensional spaces. Although we are not yet able to evaluate the two quantities r(2, 2) and $r_s(2)$ exactly, we can show that

$$\frac{4}{3} < r(2,2) \leqslant \sqrt{2}$$
, (77)

$$\frac{4}{3} < r_s(2) \leqslant \sqrt{3} \,. \tag{78}$$

To these inequalities we have to add the obvious fact that $r(2, 2) \leq r_s(2)$. To justify the lower bound in (77) (and hence also that in (78)), we observe that combining (58) and (68) yields $\rho(X, Y) \geq 4/3$ for all 2-dimensional spaces X, Y. Equality is possible iff d(X, Y) = 3/2, which happens iff the unit balls of X and Y are simultaneous linear images of a square and a regular hexagon. Without loss of generality, this is the same as saying that X is isomorphic to ℓ_1^2 . By Proposition 15, this ensures that $\rho(X, Y) =$ $\rho(\ell_1^2, Y) \geq \sqrt{2} > 4/3$. Hence, it must be the case that r(2, 2) > 4/3 strictly.

As we have already seen, the upper bound in (77) can by found by evaluating $\rho(\ell_1^2, \ell_2^2) = \sqrt{2}$, which is a special case of (59). The upper bound in (78), instead, is obtained by considering a space whose unit ball is a cleverly chosen octagon [4, Appendix D].

3.4. An important special case: quantum theory. In this section we will study the case where both parties are described by a quantum model. Before we start, let us expound some notation that we already partially introduced. We denote by M_k^{sa} the real vector space of $k \times k$ Hermitian matrices. By equipping it with a Schatten norm $\|\cdot\|_p$, defined by $\|z\|_p := (\text{tr } |z|^p)^{1/p}$, we can make such space a Banach space, which we denote by $S_p^{k,sa}$. In what follows, we will be interested in the two particular cases p = 1 and $p = \infty$, whose corresponding norms are the trace norm and the operator norm, respectively. For simplicity, we will make the canonical identification $(S_1^{k,sa})^* = S_{\infty}^{k,sa}$. Accordingly, the action of $(S_1^{k,sa})^*$ on $S_1^{k,sa}$ is given simply by the Hilbert–Schmidt inner product, i.e. y(x) = tr[xy] for $x \in S_1^{k,sa}$ and $y \in S_{\infty}^{k,sa}$. As for the tensor product, remember that $M_n^{sa} \otimes M_m^{sa} = M_{nm}^{sa}$ canonically. We now proceed to prove Theorem 8, hence determining the scaling of the function $\rho(S_1^{n,sa}, S_1^{m,sa})$ with respect to *n* and *m*. The proof of the Corollary 9 appears at the end of the present section.

Note. From now on, in some of the proofs we will find it convenient to adopt Dirac's notation for vectors and functionals in (or acting on) \mathbf{R}^n and \mathbf{C}^n . This will be done without further comments.

Proof of Theorem 8. In order to establish (38), we have to show the existence of two constants c, C > 0 such that

$$c\min\{n,m\}^{3/2} \leqslant \sup_{0 \neq z \in \mathsf{M}_n^{\mathrm{sa}} \otimes \mathsf{M}_m^{\mathrm{sa}}} \frac{\|z\|_{S_1^{n,\mathrm{sa}} \otimes \pi S_1^{m,\mathrm{sa}}}}{\|z\|_{S_1^{n,\mathrm{sa}} \otimes \varepsilon S_1^{m,\mathrm{sa}}}} \leqslant C\min\{n,m\}^{3/2}.$$
 (79)

We break down the argument to prove (79) into two parts.

Step 1: lower bound. We assume without loss of generality that $n \leq m$, and consider two Hilbert–Schmidt orthonormal bases $(x_i)_{1 \leq i \leq n^2}$ and $(y_j)_{1 \leq j \leq m^2}$ of M_n^{sa} and M_m^{sa} , respectively. We form the random tensor

$$z = \sum_{i=1}^{n^2} \sum_{j=1}^{m^2} g_{ij} x_i \otimes y_j,$$
(80)

where (g_{ij}) are independent N(0, 1) Gaussian random variables. Let us observe that the distribution of z does not depend on the choice of the local orthonormal bases. We use the results from Corollary C.4:

$$\mathbf{E} \| z \|_{S_1^{n,\mathrm{sa}} \otimes_{\varepsilon} S_1^{m,\mathrm{sa}}} \leqslant C \sqrt{n} \, m^{3/2},\tag{81}$$

$$\mathbf{E} \| z \|_{S^{n,\mathrm{sa}}_{\infty} \otimes_{\varepsilon} S^{m,\mathrm{sa}}_{\infty}} \leqslant C' \sqrt{m}.$$
(82)

By duality, (82) implies a lower bound on the projective norm of z in $S_1^{n,sa} \otimes_{\pi} S_1^{m,sa}$. More precisely, using the duality between $S_1^{n,sa} \otimes_{\pi} S_1^{m,sa}$ and $S_{\infty}^{n,sa} \otimes_{\varepsilon} S_{\infty}^{m,sa}$ together with the Cauchy–Schwartz inequality, we obtain

$$\mathbf{E}\left(\sum_{i,j}g_{ij}^{2}\right)^{1/2} = \mathbf{E}\sqrt{\mathrm{tr}[z^{2}]} \leqslant \sqrt{\mathbf{E}\,\|z\|_{S_{1}^{n,\mathrm{sa}}\otimes_{\pi}S_{1}^{m,\mathrm{sa}}}}\sqrt{\mathbf{E}\,\|z\|_{S_{\infty}^{n,\mathrm{sa}}\otimes_{\varepsilon}S_{\infty}^{m,\mathrm{sa}}}}.$$
(83)

Since the l.h.s. of (83) is of order *nm*, combining (83) and (82) yields the lower bound

$$\mathbf{E} \| z \|_{S_1^{n,\mathrm{sa}} \otimes_\pi S_1^{m,\mathrm{sa}}} \geqslant c n^2 m^{3/2}.$$
(84)

Using the above relation together with (81), we see that the random variable

$$U := C \|z\|_{S_1^{n, \mathrm{sa}} \otimes_\pi S_1^{m, \mathrm{sa}}} - cn^{3/2} \|z\|_{S_1^{n, \mathrm{sa}} \otimes_\varepsilon S_1^{m, \mathrm{sa}}}$$

has a nonnegative expectation. In particular, the event $\{U \ge 0\}$ is nonempty, from which it follows that $\rho(S_1^{n,sa}, S_1^{m,sa}) \ge cC^{-1}n^{3/2}$.

Step 2: upper bound. As before, we assume, without loss of generality, that $n \leq m$. Let us consider an element $z \in M_n^{sa} \otimes M_m^{sa}$ such that $||z||_{S_1^{n,sa} \otimes_{\varepsilon} S_1^{m,sa}} = 1$. By Corollary C.12, there is a state φ such that

$$\|\widetilde{z}(x)\|_{1} \leq 2\sqrt{2} \left(\varphi(x^{2})\right)^{1/2}$$
(85)

for every $x \in M_n^{sa}$ (the notation \tilde{z} was introduced in (42)). In Dirac's notation, the spectral decomposition of φ reads

$$\varphi = \sum_{j=1}^n \lambda_j |u_j\rangle \langle u_j| ,$$

where $(|u_j\rangle)_j$ is an orthonormal basis of \mathbb{C}^n , and $(\lambda_j)_j$ is a probability distribution. Then, it is clear that $E_{jk} := |u_j\rangle\langle u_k|$, with $j, k = 1, \dots, n$, defines a Hilbert–Schmidt orthonormal basis of the space of $n \times n$ complex matrices. Using that $(E_{jk})_{jk}, (E_{kj})_{jk}$ is a biorthogonal system, we can write

$$z = \sum_{j,k=1}^{n} E_{kj} \otimes \widetilde{z}(E_{jk}) \in S_1^n \otimes S_1^m \,.$$

If we define the Hermitian matrices $F_{jk} := E_{jk} + E_{kj}$ and $H_{jk} := i(E_{jk} - E_{kj})$, one can easily check that $F_{jk} \otimes \tilde{z}(F_{jk}) + H_{jk} \otimes \tilde{z}(H_{jk}) = 2[E_{jk} \otimes \tilde{z}(E_{kj}) + E_{kj} \otimes \tilde{z}(E_{jk})]$ and therefore

$$z = \sum_{j} E_{jj} \otimes \widetilde{z}(E_{jj}) + \frac{1}{2} \sum_{j < k} \left(F_{jk} \otimes \widetilde{z}(F_{jk}) + H_{jk} \otimes \widetilde{z}(H_{jk}) \right), \tag{86}$$

where all indices range from 1 to *n*. We then obtain the following:

$$\begin{aligned} \|z\|_{S_{1}^{n,\mathrm{sa}}\otimes_{\pi}S_{1}^{m,\mathrm{sa}}} &\stackrel{1}{\leqslant} \sum_{j} \|E_{jj}\|_{1} \|\widetilde{z}(E_{jj})\|_{1} + \frac{1}{2} \sum_{j < k} \left(\|F_{jk}\|_{1} \|\widetilde{z}(F_{jk})\|_{1} + \|H_{jk}\|_{1} \|\widetilde{z}(H_{jk})\|_{1} \right) \\ &\stackrel{2}{\leqslant} \sum_{j} \|\widetilde{z}(E_{jj})\|_{1} + \sum_{j < k} \left(\|\widetilde{z}(F_{jk})\|_{1} + \|\widetilde{z}(H_{jk})\|_{1} \right) \\ &\stackrel{3}{\leqslant} 2\sqrt{2} \sum_{j} \sqrt{\varphi(E_{jj})^{2}} + 2\sqrt{2} \sum_{j < k} \left(\sqrt{\varphi(F_{jk}^{2})} + \sqrt{\varphi(H_{jk}^{2})} \right) \\ &\stackrel{4}{=} 2\sqrt{2} \sum_{j} \sqrt{\lambda_{j}} + 4\sqrt{2} \sum_{j < k} \sqrt{\lambda_{j} + \lambda_{k}} \\ &\stackrel{5}{\leqslant} 2\sqrt{2} \left(\sqrt{\pi} \left(\sum_{j} \lambda_{j} \right)^{1/2} + 2\sqrt{\frac{n(n-1)}{2}} \left(\sum_{j < k} (\lambda_{j} + \lambda_{k}) \right)^{1/2} \right) \end{aligned}$$

$$\stackrel{6}{=} 2\sqrt{2} \left(\sqrt{n} + 2\sqrt{\frac{n(n-1)}{2}} \sqrt{n-1} \right)$$
$$= 4n^{3/2} - 2\sqrt{2}(\sqrt{2}-1)\sqrt{n} \,.$$

The justification of the above steps is as follows: 1: we used the decomposition (86) as an ansatz into the minimisation that defines the projective norm (11); 2: we observed that $||E_{jj}||_1 = 1$ and $||F_{jk}||_1 = 2 = ||H_{jk}||_1$ for all j < k; 3: follows from (85); 4: we evaluated $\varphi(E_{jj}^2) = \varphi(E_{jj}) = \lambda_j$ and

$$\varphi(F_{jk}^2) = \varphi(H_{jk}^2) = \varphi(|u_j\rangle\langle u_j| + |u_k\rangle\langle u_k|) = \lambda_j + \lambda_k;$$

5: is an application of the Cauchy-Schwartz inequality; 6: we computed

$$\sum_{j < k} (\lambda_j + \lambda_k) = \frac{1}{2} \sum_{j \neq k} (\lambda_j + \lambda_k) = \frac{1}{2} \left(\sum_j (n-1)\lambda_j + \sum_k (n-1)\lambda_k \right) = n-1,$$

and remembered that $\sum_{j} \lambda_j = 1$. This completes the proof of (39), which in turn implies (38) with C = 4. \Box

Proof of Corollary 9. The claim (40) derives from (79). Indeed: (i) the norm $\|\cdot\|_{LO}$ as defined by [52] satisfies $\|\cdot\|_{LO} \leq \|\cdot\|_1$ e.g. by Helstrom's theorem; (ii) the inequality $\|\cdot\|_1 \leq \|\cdot\|_{S_1^{n,sa} \otimes_{\pi} S_1^{m,sa}}$ follows from (12) combined with the fact that the standard quantum mechanical composition rule yields a legitimate composite in the GPT sense; and (iii) $\|\cdot\|_{LO} \geq \|\cdot\|_{S_1^{n,sa} \otimes_{\varepsilon} S_1^{m,sa}}$ by [32, Proposition 22]. \Box

4. Asymptotic Lower Bounds on the π/ε Ratio

4.1. A lower bound for two copies of the same theory. In this section we prove that $r_s(n) \ge c\sqrt{n}/\log^3 n$ for a certain universal constant c > 0, which is the technically challenging part of the statement of Theorem 4. In fact, remember that the example of ℓ_1^n shows that $r_s(n) \le \sqrt{2n}$ (as reported in (20)), hence the aforementioned result is optimal up to logarithmic factors. For an *n*-dimensional Banach space we denote $d_X = d(X, \ell_2^n)$, where d is the Banach–Mazur distance.

Theorem 17. There exists a universal constant c > 0 such that for every Banach space *X* of dimension *n*, we have

$$\rho(X, X) \ge \frac{cn}{d_X \log^3 n} \,.$$

In particular, $r_s(n) \ge c\sqrt{n}/\log^3 n$.

Proof. The lower bound on $r_s(n)$ follows immediately by combining (17) and the well-known estimate $d_X \leq \sqrt{n}$ in (54). We now set out to prove (17). We may assume that X is equal to $(\mathbb{R}^n, \|\cdot\|_X)$, with

$$\|\operatorname{Id}: \ell_2^n \to X\| \cdot \|\operatorname{Id}: X \to \ell_2^n\| = d_X.$$
(87)

Our main tool is the following lemma, whose proof we postpone. It is based on ideas from [49] (see also [53, Lemma 46.2] and comments below it). We point that the assumption on the norm is not a restriction: since $d_X \leq \sqrt{n}$, X is isometric to $(\mathbf{R}^n, \|\cdot\|)$ for a norm $\|\cdot\|$ on \mathbf{R}^n satisfying $\frac{1}{\sqrt{n}} |\cdot| \leq \|\cdot\|_X \leq |\cdot|$.

Lemma 18. Consider a Banach space $X = (\mathbf{R}^n, \|\cdot\|_X)$, and assume that $\frac{1}{\sqrt{n}}|\cdot| \leq \|\cdot\|_X \leq |\cdot|$, where $|\cdot|$ is the standard Euclidean norm. Then there exist orthonormal vectors $|f_i\rangle \in \mathbf{R}^n$, $i = 1, \dots, k$, with $k \geq cn/\log(n)$, such that

$$\mathbf{E} \left\| \sum_{i=1}^{k} g_i \left| f_i \right\rangle \right\|_{X} \mathbf{E} \left\| \sum_{i=1}^{k} g_i \left\langle f_i \right| \right\|_{X^*} \leqslant Cn \log n \,.$$
(88)

Here, c and C are universal constants and $(g_i)_{i=1}^n$ is a sequence of independent N(0, 1) Gaussian random variables.

Let us consider the vectors $(|f_i\rangle)_{1 \le i \le k} \in \mathbb{R}^n$ from Lemma 18, and form the random tensors $|z\rangle = \sum_{i,j=1}^k g_{ij} |f_i\rangle \otimes |f_j\rangle \in X \otimes X$ and $\langle z| = \sum_{i,j=1}^k g_{ij} \langle f_i| \otimes \langle f_j| \in X^* \otimes X^*$, where (g_{ij}) are independent N(0, 1) Gaussian random variables. It is clear that

$$\mathbf{E}\langle z|z\rangle = \mathbf{E}\sum_{i,j=1}^{k} g_{ij}^2 = k^2 \geqslant \frac{c^2 n^2}{\log^2 n}.$$
(89)

On the other hand, according to Chevet's inequality (Theorem 3), we have

$$\mathbf{E} \||z\rangle\|_{X\otimes_{\varepsilon} X} = \mathbf{E} \left\| \sum_{i,j=1}^{k} g_{ij} |f_i\rangle \otimes |f_j\rangle \right\|_{X\otimes_{\varepsilon} X} \leqslant 2 \mathbf{E} \left\| \sum_{i=1}^{k} g_i |f_i\rangle \right\|_{X} \|\mathrm{Id}_K : \ell_2^k \to X\|,$$
(90)

$$\mathbf{E} \|\langle z \| \|_{X^* \otimes_{\varepsilon} X^*} = \mathbf{E} \left\| \sum_{i,j=1}^k g_{ij} \langle f_i | \otimes \langle f_j | \right\|_{X^* \otimes_{\varepsilon} X^*} \\ \leqslant 2 \mathbf{E} \left\| \sum_{i=1}^k g_i \langle f_i | \right\|_{X^*} \| \mathrm{Id}_K : \ell_2^k \to X^* \|,$$
(91)

where Id_K denotes the identity map restricted to $K := span\{|f_i\rangle : 1 \le i \le k\}$. We can then write

$$\begin{split} \mathbf{E}\langle z|z\rangle &\stackrel{1}{\leqslant} \mathbf{E} \left\| \sum_{i,j=1}^{k} g_{ij} |f_i\rangle \otimes |f_j\rangle \right\|_{X\otimes_{\pi} X} \left\| \sum_{i,j=1}^{k} g_{ij} \langle f_i| \otimes \langle f_j| \right\|_{X^*\otimes_{\varepsilon} X^*} \\ &\stackrel{2}{\leqslant} \rho(X,X) \mathbf{E} \left\| \sum_{i,j=1}^{k} g_{ij} |f_i\rangle \otimes |f_j\rangle \right\|_{X\otimes_{\varepsilon} X} \left\| \sum_{i,j=1}^{k} g_{ij} \langle f_i| \otimes \langle f_j| \right\|_{X^*\otimes_{\varepsilon} X^*} \\ &\stackrel{3}{\leqslant} \rho(X,X) \left(\mathbf{E} \left\| \sum_{i,j=1}^{k} g_{ij} |f_i\rangle \otimes |f_j\rangle \right\|_{X\otimes_{\varepsilon} X}^2 \right)^{\frac{1}{2}} \\ & \left(\mathbf{E} \left\| \sum_{i,j=1}^{k} g_{ij} \langle f_i| \otimes \langle f_j| \right\|_{X^*\otimes_{\varepsilon} X^*}^2 \right)^{\frac{1}{2}} \end{split}$$

$$\begin{split} \stackrel{4}{\leqslant} C_2^2 \rho(X, X) & \mathbf{E} \left\| \sum_{i,j=1}^k g_{ij} \left| f_i \right\rangle \otimes \left| f_j \right\rangle \right\|_{X \otimes_{\varepsilon} X} \mathbf{E} \left\| \sum_{i,j=1}^k g_{ij} \left\langle f_i \right| \otimes \left\langle f_j \right| \right\|_{X^* \otimes_{\varepsilon} X^*} \\ \stackrel{5}{\leqslant} 4C_2^2 \rho(X, X) & \mathbf{E} \left\| \sum_{i=1}^k g_i \left| f_i \right\rangle \right\|_X \mathbf{E} \left\| \sum_{i=1}^k g_i \left\langle f_i \right| \right\|_{X^*} \\ & \| \mathrm{Id}_K : \ell_2^k \to X \| \cdot \| \mathrm{Id}_K : \ell_2^k \to X^* \| \\ \stackrel{6}{\leqslant} 4C_2^2 \rho(X, X) d_X & \mathbf{E} \left\| \sum_{i=1}^k g_i \left| f_i \right\rangle \right\|_X \mathbf{E} \left\| \sum_{i=1}^k g_i \left\langle f_i \right| \right\|_{X^*} \\ \stackrel{7}{\leqslant} C' \rho(X, X) d_X n \log n \,. \end{split}$$

The justification of the above steps is as follows: 1: we used the duality of injective and projective norm (44); 2: follows by definition of $\rho(X, X)$; 3: is an application of the Cauchy–Schwarz inequality; 4: is the p = 2 case of the Khintchine–Kahane inequality (Theorem 2); 5: derives from (90) and (91); 6: can be derived from (87), using the fact that orthogonal projections onto subspaces of Hilbert spaces have norm 1; 7: is the statement of Lemma 18. Combining the above estimate with the lower bound in (89), we deduce that

$$\rho(X, X) \ge c \, \frac{n}{d_X \log^3 n} \, ,$$

which concludes the proof.

Proof of Lemma 18. According to the MM^* -estimate (Theorem C.1), there exists an isomorphism $T : \ell_2^n \to X$ such that

$$\ell_X(T)\ell_{X^*}((T^{-1})^*) \leqslant Cn\log n.$$
(92)

Moreover, since $\ell_X(T) = \ell_X(T \circ U)$ for every unitary U, it can be assumed that T is positive definite. By the spectral theorem, T can be written as

$$T = \sum_{i=1}^{n} \lambda_i |f_i\rangle \langle f_i| ,$$

for some positive numbers λ_i and $(|f_i\rangle)_{i=1}^n$ an orthonormal basis of \mathbb{R}^n . Then, inequality (92) implies that

$$\mathbf{E}\left\|\sum_{i=1}^{n}\lambda_{i}g_{i}|f_{i}\rangle\right\|_{X}\mathbf{E}\left\|\sum_{i=1}^{n}\lambda_{i}^{-1}g_{i}\langle f_{i}|\right\|_{X^{*}} \leq Cn\log n$$

Using the inequalities $|||f_i\rangle||_X \ge 1/\sqrt{n}$ and $||\langle f_j|||_{X^*} \ge 1$ together with Jensen inequality (or Lemma C.5), we see that $\lambda_i \lambda_j^{-1} \le Cn^{3/2} \log n$ holds for any indices *i*, *j*. Let us denote $m = \min\{\lambda_k : 1 \le k \le n\}$ and $M = \max\{\lambda_k, 1 \le k \le n\}$. It follows that

$$\frac{M}{m} \leqslant C n^{3/2} \log n \,. \tag{93}$$

Now, (93) implies that the sets

$$A_s = \left\{ 1 \leqslant j \leqslant n : 2^{s-1} \leqslant \frac{\lambda_j}{m} \leqslant 2^s \right\}$$

with $s = 1, \dots, r$, define a partition of $\{1, \dots, n\}$ for a certain $r \leq C' \log n$. By the pigeonhole principle, one can immediately deduce the existence of a set A_{s_0} such that $|A_{s_0}| \geq \frac{n}{C' \log n}$. Now, consider the set of orthonormal vectors $\{|f_j\rangle: j \in A_{s_0}\}$. Applying Lemma C.5, we see that

$$\begin{split} \mathbf{E} \left\| \sum_{i \in A_{s_0}} g_i |f_i\rangle \right\|_X \mathbf{E} \left\| \sum_{i \in A_{s_0}} g_i \langle f_i | \right\|_{X^*} \\ \leqslant \mathbf{E} \left\| \sum_{i \in A_{s_0}} \frac{\lambda_i}{2^{s_0 - 1} m} g_i |f_i\rangle \right\|_X \mathbf{E} \left\| \sum_{i \in A_{s_0}} 2^{s_0} m \lambda_i^{-1} g_i \langle f_i | \right\|_{X^*} \\ \leqslant 2 \mathbf{E} \left\| \sum_{i \in A_{s_0}} \lambda_i g_i |f_i\rangle \right\|_X \mathbf{E} \left\| \sum_{i \in A_{s_0}} \lambda_i^{-1} g_i \langle f_i | \right\|_{X^*} \\ \leqslant Cn \log n \,, \end{split}$$

completing the argument. \Box

4.2. A lower bound for any pair of theories. The aim of this section is to prove Theorems 5 and 6, which provide general lower bounds on $\rho(X, Y)$ as functions of the dimensions n, m only. As discussed in Sect. 2, our strategy requires two preliminary results that allow to reduce the problem to the more manageable special case where either X or Y is one of the 'classical' spaces ℓ_1^n , ℓ_2^n or ℓ_{∞}^n . We start by presenting the solution to these special cases.

Lemma 19. For every finite-dimensional normed space X with dim $X \ge n$, we have

(a) $\rho(\ell_1^n, X) \ge \sqrt{n/2}$, (b) $\rho(\ell_2^n, X) \ge \sqrt{n}$, (c) $\rho(\ell_\infty^n, X) \ge \sqrt{n/2}$.

Proof. We already know from Proposition 13 that $\rho(\ell_1^n, X) = \pi_1^{(n)}(\mathrm{Id}_X)$. Moreover (this is especially clear from (73)), we have $\rho(\ell_1^n, Y) \leq \rho(\ell_1^n, X)$ whenever Y is a subspace of X. Consequently, it suffices to prove (a) in the case when dim X = n. In that case, we argue that

$$\pi_1^{(n)}(\mathrm{Id}_X) \geqslant \pi_2^{(n)}(\mathrm{Id}_X) \geqslant \frac{1}{\sqrt{2}}\pi_2(\mathrm{Id}_X) \geqslant \sqrt{n/2},$$

where we used points 1., 2. and 3. from Proposition C.10.

Part (b) is a direct consequence of Proposition C.8 together with the formulation of $\rho(\ell_2^n, X)$ from Lemma 11.

Finally, (c) follows from (a) since $\rho(\ell_{\infty}^n, X) = \rho(\ell_1^n, X^*)$, cf (61).

Remark. In light of the discussion at the end of Sect. 1.3 (see also the Remark after Proposition 15), we see that Lemma 19 entails the following: the gap between local and global bias for XOR games played over any system *AB* in which e.g. $A = \text{Cl}_n$ is a classical theory (defined in (7)) is at least $\sqrt{n/2}$ whenever dim $B \ge n$.

The following result is a variant of the $\ell_1/\ell_2/\ell_{\infty}$ trichotomy' which is based on ideas from Pisier [48], Rudelson [7], and Szarek–Tomczak-Jaegerman [50].

Theorem 20. Let X be a normed space of dimension n. Then for every $1 \le A \le \sqrt{n}$ at least one of the following holds

- 1. *X* contains a subspace of dimension $d := c\sqrt{n}$ which is $CA\sqrt{\log n}$ -isomorphic to ℓ_{∞}^d .
- 2. X^* contains a subspace of dimension d which is $CA\sqrt{\log n}$ -isomorphic to ℓ_{∞}^d .
- 3. *X* contains a C log n-complemented 4-Euclidean subspace of dimension $cA^2/\log n$.

Here, C and c are universal constants.

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Proof. By the *MM**-estimate (Theorem C.1), we may assume that $X = (\mathbf{R}^n, \|\cdot\|_X)$ with

$$\ell_X(\mathrm{Id}) \leqslant C\sqrt{n\log n}$$
 and $\ell_{X^*}(\mathrm{Id}) \leqslant C\sqrt{n\log n}$.

Let \mathcal{E} be the John ellipsoid of X as defined in Theorem C.7, $(|e_i\rangle)_{1 \le i \le n}$ be the semiaxes of \mathcal{E} and (λ_i) their lengths, i.e. $\mathcal{E} = T(B_2^n)$ where $T = \sum \lambda_i |e_i\rangle\langle e_i|$. Assume also that $\lambda_1 \le \lambda_2 \le \cdots \le \lambda_n$. Note that we can assume that T is of this form because $T \circ u$ defines the same ellipsoid for every orthogonal transformation u. We consider the following dichotomy.

Case (i) $\lambda_{n/3} \leq A/\sqrt{n}$. Let $E = \text{span}\{|e_i\rangle : 1 \leq i \leq n/3\}$ and P_E be the orthogonal projection onto E. We note that P_E is orthogonal for both the standard Euclidean structure in \mathbb{R}^n and the Euclidean structure induced by \mathcal{E} (i.e. using $(\lambda_i |e_i\rangle)$) as an orthonormal basis). We apply Theorem C.9 to P_E in order to produce an m-dimensional subspace of X which is R-isomorphic to ℓ_{∞}^m , for $m = c\sqrt{n}$ and $R = C\ell'_X(P_E)$, where we denote by ℓ'_X the ℓ_X -norm computed using the Euclidean structure induced by \mathcal{E} . We use Lemma C.5 to obtain the bound

$$\ell'_X(P_E) = \mathbf{E} \left\| \sum_{i=1}^{n/3} g_i \lambda_i |e_i\rangle \right\|_X \leqslant \lambda_{n/3} \ell_X(P_E) \leqslant \frac{A}{\sqrt{n}} \ell_X(\mathrm{Id}) \leqslant CA\sqrt{\log n},$$

and we conclude that X contains a subspace which is $CA\sqrt{\log n}$ -isomorphic to $\ell_{\infty}^{c\sqrt{n}}$.

Case (ii) $\lambda_{n/3} > A/\sqrt{n}$. Let $F = \text{span}\{e_i : i > n/3\}$ and denote by $\text{Id}_F : F \to \mathbb{R}^n$ the identity map restricted to F. We have

$$\|\mathrm{Id}_F:\ell_2^n\to X\|\leqslant \frac{\sqrt{n}}{A}.$$

To see the previous bound, just write $\mathrm{Id}_F = (T \circ T^{-1}|_F)$ and use that $||T^{-1}|_F : \ell_2^n \to \ell_2^n || \leq \sqrt{n}/A$ and $||T : \ell_2^n \to X|| \leq 1$.

We apply the same dichotomy to X^* . If case (i) occurs for either X or X^* , we are done. It remains to consider the situation when case (ii) occurs for both. This means that there exist subspaces F_1 and F_2 of dimension 2n/3 such that

$$\|\operatorname{Id}_{F_1} : \ell_2^n \to X\| \leqslant \sqrt{n}/A, \|\operatorname{Id}_{F_2} : \ell_2^n \to X^*\| \leqslant \sqrt{n}/A.$$

Consider the subspace $F = F_1 \cap F_2$ (note that dim $F \ge n/3$). We are going to apply the Dvoretzky–Milman theorem (Theorem C.6) to both $X \cap F$ and $X^* \cap F$ (that is, to the

space F with the norms inherited from X and from X^* respectively). The corresponding Dvoretzky dimensions are

$$k_{*}(X \cap F) = \left(\frac{\ell_{X \cap F}(\mathrm{Id}_{F})}{\|\mathrm{Id}_{F} : \ell_{2}^{n} \to X\|}\right)^{2} \ge \frac{A^{2}}{n} \ell_{X \cap F}(\mathrm{Id}_{F})^{2} = A^{2} \ell_{X}(P_{F})^{2}/n,$$

$$k_{*}(X^{*} \cap F) = \left(\frac{\ell_{X^{*} \cap F}(\mathrm{Id}_{F})}{\|\mathrm{Id}_{F} : \ell_{2}^{n} \to X^{*}\|}\right)^{2} \ge \frac{A^{2}}{n} \ell_{X^{*} \cap F}(\mathrm{Id}_{F}) = A^{2} \ell_{X^{*}}(P_{F})^{2}/n.$$

On the other hand, if we consider the random vector $g = \sum_{i=1}^{\dim F} g_i f_i$, where (g_i) are independent N(0, 1) Gaussian random variables and (f_i) is an orthonormal basis in F, we compute

$$\frac{n}{3} \leq \dim(F) = \mathbf{E} |g|^2 \stackrel{1}{\leq} \left(\mathbf{E} ||g||_X^2 \right)^{1/2} \left(\mathbf{E} ||g||_{X^*}^2 \right)^{1/2}$$
$$\stackrel{2}{\leq} C \mathbf{E} ||g||_X \mathbf{E} ||g||_{X^*} = C\ell_X(P_F)\ell_{X^*}(P_F),$$

where 1 follows from Cauchy–Schwarz inequality together with the inequality $|g|^2 \leq ||g||_X ||g||_{X^*}$, and 2 from Khintchine–Kahane inequalities (Theorem 2). Since $\ell_{X^*}(P_F) \leq \ell_{X^*}(\mathrm{Id}) \leq C\sqrt{n \log n}$, we have $\ell_X(P_F) \geq c\sqrt{n/\log n}$, and similarly $\ell_{X^*}(P_F) \geq c\sqrt{n/\log n}$. It follows that

$$k_*(X \cap F) \ge cA^2/\log n$$
 and $k_*(X^* \cap F) \ge cA^2/\log n$.

By Dvoretzky–Milman theorem (Theorem C.6) and the remark following it, there is a subspace $E \subseteq F$ of dimension $cA^2/\log n$ such that both $X \cap E$ and $X^* \cap E$ are 4-Euclidean. Moreover, using the extra information given by Theorem C.6, we have

$$\|\mathrm{Id}_E : \ell_2^n \to X\| \leqslant \frac{2\ell_X(P_F)}{\sqrt{n/3}},$$
$$\|\mathrm{Id}_E : \ell_2^n \to X^*\| \leqslant \frac{2\ell_X(P_F)}{\sqrt{n/3}}.$$

Since $||P_E : X \to \ell_2^n|| = ||P_E : \ell_2^n \to X^*|| = ||Id_E : \ell_2^n \to X^*||$, we have

$$\begin{aligned} \|P_E : X \to X\| &\leq \|P_E : X \to \ell_2^n\| \cdot \|P_E : \ell_2^n \to X\| \\ &\leq \frac{C\ell_X(P_F)\ell_{X^*}(P_F)}{n} \\ &\leq \frac{C\ell_X(\mathrm{Id})\ell_{X^*}(\mathrm{Id})}{n} \\ &\leq C' \log n \end{aligned}$$

and therefore E is a $C' \log n$ -complemented 4-Euclidean subspace of X.

Proof of Theorem 5. As usual, the difficult part is to establish the lower bound on r(n, n), while the upper bound is reported in (19) (set n = m). Let X and Y be n-dimensional normed spaces, and A > 1 be a number whose value will be optimised later. Theorem 20 implies in particular that at least one of the following occurs (here we use the classical fact that a subspace isometric to ℓ_{∞}^m is automatically 1-complemented):

(i)
$$f\left(\ell_{\infty}^{c\sqrt{n}}, X\right) \leq CA\sqrt{\log n};$$

(ii) $f\left(\ell_1^{c\sqrt{n}}, X\right) \leq CA\sqrt{\log n}$; or (iii) $f\left(\ell_2^{cA^2/\log n}, X\right) \leq C\log n$.

If (i) holds, then by Lemma 19 and (64) we obtain

$$\rho(X,Y) \ge \rho\left(\ell_{\infty}^{c\sqrt{n}},Y\right) / f\left(\ell_{\infty}^{c\sqrt{n}},X\right) \ge \frac{cn^{1/4}}{A\sqrt{\log n}}$$

A similar estimate applies when X satisfies (ii), or when Y satisfies (i) or (ii). The only uncovered case is when X and Y both satisfy (iii), and we have then

$$\rho(X,Y) \ge \rho\left(\ell_2^{cA^2/\log n}, \ell_2^{cA^2/\log n}\right) / (C\log n)^2 \ge \frac{cA^2}{\log^3 n}$$

The optimal choice is $A = n^{1/12} (\log n)^{5/6}$, which gives the announced lower bound. \Box

Proof of Theorem 6. The upper bound on r(n, m) follows again from (19), so we focus on the lower bound. Let *X* and *Y* be normed spaces of respective dimensions *n* and *m* (with $n \le m$), and A > 1 be a number whose value will be optimised later. As is the previous proof, we combine Theorem 20 (applied only to *X*) and Lemma 19. In case (i), we have

$$\rho(X,Y) \ge \rho\left(\ell_{\infty}^{c\sqrt{n}},Y\right) / f\left(\ell_{\infty}^{c\sqrt{n}},X\right) \ge \frac{cn^{1/4}}{A\sqrt{\log n}}.$$

Case (ii) is similar by duality. In case (iii), we have

$$\rho(X,Y) \ge \rho\left(\ell_2^{cA^2/\log n},Y\right) / f\left(\ell_2^{cA^2/\log n},X\right) \ge \frac{cA}{(\log n)^{3/2}}$$

The optimal choice $A = n^{1/8} \sqrt{\log n}$ always gives the lower bound $\rho(X, Y) \ge cn^{1/8} / \log n$, concluding the proof. \Box

5. Conclusions

In this paper, we have defined and investigated XOR games from the foundational standpoint provided by general probabilistic theories. This has led us to identify a deep connection between the minimum relative increase in the bias when global strategies displace local ones on the one hand, and the so-called projective/injective ratio on the other. The existence of such a connection is made possible by the fact that all norms on a given vector space can be well approximated by suitable base norms induced by GPTs.

The projective/injective ratio r(n, m) is a universal function over pairs of integers that encodes some structural information about the theory of Banach spaces. For instance, we have shown that n/r(n, n) provides a lower bound on the diameter of the Banach– Mazur compactum in dimension n as measured by the weak distance [6,7]. We have also proved that r(n, m) is always lower bounded by a universal constant strictly larger than 1. This shows the remarkable fact that injective and projective tensor product can never be isometric, even though Pisier's celebrated construction [8] demonstrates that they can be isomorphic in the case where the spaces have infinite dimension. Along the way, we developed an Auerbach–type lemma that may be of independent interest. The main results we have presented concern the asymptotic behaviour of the ratio r(n, m) and of its symmetrised version $r_s(n)$. In this context, we were able to show that, up to logarithmic factors, $r_s(n)$ is of the order \sqrt{n} . We showed that r(n, m) grows at least as min $\{n, m\}^{1/8}$, and that one can improve the exponent to 1/6 if n = m. The proofs of these latter results follow by putting together an understanding of the projective/injective ratio in tensor products of the form $\ell_p^d \otimes X$, with $p = 1, 2, \infty$, and a 'trichotomy theorem' that identifies in any normed space a sufficiently large subspace that is close in the Banach–Mazur distance to either (a) ℓ_1^d , or (b) ℓ_2^d , or (c) ℓ_{∞}^d . The main technical hurdle consists in establishing the additional requirement that in case (b) the chosen subspace is also well-complemented. As we have discussed, our findings draw on previous ideas by Pisier, Rudelson, Szarek, and Tomczak-Jaegermann.

Finally, although our primary subject of study is the intrinsic theory of XOR games played over general physical systems, it would be futile to deny that quantum systems hold great importance, due to their omnipresence in Nature as we currently understand it. In this spirit, we determined the exact scaling of the maximal global/local bias ratio in quantum XOR games, finding that it is of the order min $\{n, m\}^{3/2}$, with *n*, *m* being the local Hilbert space dimensions. Interestingly, this implies a new bound on the maximal strength of quantum mechanical data hiding against local operations.

Our work leaves a number of open problems that we believe are worth investigating. Let us recall briefly some of them. First, it would be interesting to compute exactly the absolute minimum of r(n, m) across all pairs of integers, which one may conjecture to be equal to $\sqrt{2}$. A perhaps more profound question is to determine the best exponent γ_{opt} such that $r(n, m) \ge c \min\{n, m\}^{\gamma_{opt}}$ for all n, m. We ask whether $\gamma_{opt} = 1/2$. As we have seen, the simplified statement with n = m would follow from Rudelson's conjecture [7] that the Banach–Mazur compactum in dimension n has a diameter of the order \sqrt{n} with respect to the weak distance.

Acknowledgements. Open Access funding provided by Projekt DEAL. We are grateful to Gilles Pisier for sharing with us an unpublished proof of the estimate $r_s(n) \ge n^{1/10-o(1)}$, and to Marius Junge for helpful comments on some of our results. We thank the Institut Henri Poincaré for support and hospitality during the programme 'Analysis in Quantum Information Theory' when part of the work on this paper was performed. GA was supported in part by ANR (France) under the Grant StoQ (2014-CE25-0003). LL acknowledges financial support from the European Research Council (ERC) under the Starting Grant GQCOP (Grant No. 637352). CP is partially supported by the Spanish 'Ramón y Cajal Programme' (RYC-2012-10449), the Spanish 'Severo Ochoa Programme' for Centres of Excellence (SEV-2015-0554) and the Grant DTM2014-54240-P, funded by Spanish MINECO. The research of SJS was supported in part by a Grant DMS-1600124 from the National Science Foundation (U.S.A.). AW acknowledges support from the Spanish MINECO, Project FIS2016-86681-P, with the support of FEDER funds, and the Generalitat de Catalunya, CIRIT Project 2014-SGR-966.

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Appendix A. More on XOR Games in GPTs

Throughout this appendix we will demonstrate that the connection drawn by Theorem 1 between injective norms and bias of XOR games under local strategies is in a certain sense robust. Namely, we will show that allowing the players to use a bounded amount of back-and-forth communication does not make the bias larger that a constant times the same injective norm as in (22). In other words, the bias does not grow by more than a constant factor with respect to the purely local case.

Our argument does not require the communication to be classical. In fact, in principle the players are allowed to exchange any physical system described by a GPT. For instance, Alice could initiate the protocol by manipulating the subsystem A corresponding to her share of the question so as to prepare a bipartite state of a new system A_1M_1 ; the subsystem M_1 is sent to Bob, while Alice keeps A_1 for later use; then, Bob employs M_1 together with his share of the question B to prepare a message M'_1 to be sent to Alice and a record B_1 for later use. After N such rounds, Alice will have sent the systems M_1, \ldots, M_N , and Bob will have sent the systems M'_1, \ldots, M'_N . The total dimension of the systems exchanged is thus

$$L_{\leftrightarrow} := (\dim(M_1) \dots \dim(M_N))(\dim(M_1') \dots \dim(M_N')). \tag{A1}$$

In what follows, we will refer to such a setting as a 'local strategy assisted by two-way communication of total dimension L_{\leftrightarrow} '. We now deal with the problem of bounding the corresponding bias.

Note. We will often consider complicated compositions of maps acting on different systems. The convention we adopt is to omit all occurrences of the identity map acting on the untouched systems. In this way, if $T : A \rightarrow BC$ and $S : B \rightarrow DE$ are linear maps, we write ST instead of $(S_B \otimes \text{Id}_C) T_A$.

Proposition A.1. Let $G = (AB, \omega, p, c)$ be an XOR game over a bipartite GPT AB, and set $z_G = \sum_i p_i (-1)^{c_i} \omega_i$ as in Theorem 1. The bias $\beta_{\leftrightarrow}(G)$ of G under local strategies assisted by two-way communication of total dimension L_{\leftrightarrow} can be upper bounded as

$$\beta_{\leftrightarrow}(G) \leqslant \sup_{\dim W \leqslant L_{\leftrightarrow}} \left\| \widetilde{z}_G \otimes \operatorname{Id}_W : V_A^* \otimes_{\varepsilon} W \longrightarrow V_B \otimes_{\pi} W \right\|$$
$$\leqslant L_{\leftrightarrow} \| z_G \|_{V_A \otimes_{\varepsilon} V_B} = L_{\leftrightarrow} \beta_{\operatorname{LO}}(G) ,$$
(A2)

where the optimisation is over all normed spaces W of dimension up to L_{\leftrightarrow} , and \tilde{z}_G : $V_A^* \to V_B$ is the linear map associated with the tensor $z_G \in V_A \otimes V_B$ according to (42).

Proof. The *n* rounds of communication can be represented by linear maps $T_{\alpha} : V_{A_{\alpha-1}} \otimes V_{M'_{\alpha-1}} \to V_{A_{\alpha}} \otimes V_{M_{\alpha}}$ and $S_{\alpha} : V_{B_{\alpha-1}} \otimes V_{M_{\alpha}} \to V_{B_{\alpha}} \otimes V_{M'_{\alpha}}$, for $\alpha = 1, ..., N$, where for convenience we identified $A_0 := A$, $B_0 := B$, and $V_{M'_0} := \mathbf{R}$. After the communication stage has taken place, Alice is left with the systems $A_N M'_N$, while Bob will have only B_N . They then perform local measurements to output the answers. These can be conveniently represented as $\left\{\frac{u+\varphi}{2}, \frac{u-\varphi}{2}\right\}$ (on Alice's side) and $\left\{\frac{u+\psi}{2}, \frac{u-\psi}{2}\right\}$ (on Bob's side), where $\varphi \in V_{A_N}^* \otimes V_{M'_N}^*$ and $\psi \in V_{B_N}^*$. A reasoning analogous to that in the proof of Theorem 1 shows that the bias for this strategy will be given by

$$\beta = (\varphi \otimes \psi) \left((S_N T_N) \dots (S_1 T_1)(z_G) \right) =: w(z_G) = \operatorname{tr}[\widetilde{z}_G \widetilde{w}^*], \quad (A3)$$

where we observed that the validity of the above equation for all z_G defines a functional $w \in V_A^* \otimes V_B^*$ (which depends on φ , ψ , and all the maps T_α , S_α , for $\alpha = 1, ..., N$), and for the last step we used (43).

Now, we claim that the rank of the operator $\widetilde{w}: V_A \to V_B^*$ satisfies

$$\operatorname{rk}(\widetilde{w}) \leqslant L_{\leftrightarrow} \,. \tag{A4}$$

This can be verified straightforwardly by considering for all α families of vectors $\{x_{\alpha,j_{\alpha}} \in V_{M_{\alpha}}\}_{j_{\alpha}=1,...,\dim(M_{\alpha})}, \{y_{\alpha,k_{\alpha}} \in V_{M_{\alpha}'}\}_{k_{\alpha}=1,...,\dim(M_{\alpha}')}$ and families of maps $\{\overline{T}_{\alpha,j_{\alpha}} : V_{A_{\alpha-1}} \otimes V_{M_{\alpha-1}} \rightarrow V_{A_{\alpha}}\}_{j_{\alpha}=1,...,\dim(M_{\alpha})}, \{\overline{S}_{\alpha,k_{\alpha}} : V_{B_{\alpha-1}} \otimes V_{M_{\alpha}} \rightarrow V_{B_{\alpha}}\}_{k_{\alpha}=1,...,\dim(M_{\alpha}')}$ such that one can expand

$$T_{\alpha} = \sum_{j_{\alpha}=1}^{\dim(M_{\alpha})} \overline{T}_{\alpha, j_{\alpha}} \otimes x_{\alpha, j_{\alpha}}, \qquad S_{\alpha} = \sum_{k_{\alpha}=1}^{\dim(M_{\alpha}')} \overline{S}_{\alpha, k_{\alpha}} \otimes y_{\alpha, k_{\alpha}}.$$

Defining the 'reduced' maps $\overline{T}_{\alpha, j_{\alpha}} y_{\alpha-1, k_{\alpha-1}} : V_{A_{\alpha-1}} \to V_{A_{\alpha}}$ (for $\alpha = 2, ..., N$) and $\overline{S}_{\alpha, k_{\alpha}} x_{\alpha, j_{\alpha}} : V_{B_{\alpha-1}} \to V_{B_{\alpha}}$ (for $\alpha = 1, ..., N$), we see that

$$w = \sum_{\{j_{\alpha}, k_{\alpha}\}_{\alpha}} \varphi \left(y_{N,k_{N}} \otimes (\overline{T}_{N,j_{N}} y_{N-1,k_{N-1}}) \dots (\overline{T}_{2,j_{2}} y_{1,k_{1}}) \overline{T}_{1,j_{1}} \right) \\ \otimes \psi \left((\overline{S}_{N,k_{N}} x_{N,j_{N}}) \dots (\overline{S}_{1,k_{1}} x_{1,j_{1}}) \right),$$

where the first tensor factors are functional in V_A^* , and the second belong to V_B^* . Since the above sum contains exactly L_{\leftrightarrow} terms, we see that (A4) follows. As it turns out, w satisfies also the inequality

$$\|\widetilde{w}: V_A \to V_B^*\| \leqslant 1.$$
(A5)

To see why this is the case, observe that the bias $\beta = w(z_G)$ cannot be larger than the maximal bias achievable by global strategies, as given by Theorem 1. This implies that $w(z_G) \leq ||z_G||_{V_A \otimes_{\pi} V_B}$. Since this has to hold for all $z_G \in V_A \otimes V_B$, and injective and projective tensor product are dual to each other by (44), we deduce that $1 \geq ||w||_{(V_A \otimes_{\pi} V_B)^*} = ||w||_{V_A^* \otimes_{\varepsilon} V_B^*} = ||\widetilde{w}| : V_A \to V_B^*||$, where the last equality is an application of (45).

Putting together (A3), (A4), and (A5), we see that

$$\beta_{\leftrightarrow}(G) \leqslant \sup_{\substack{w \in V_A^* \otimes V_B^* \\ \mathrm{rk}(\widetilde{w}) \leqslant L_{\leftrightarrow} \\ \|\widetilde{w}: V_A \to V_B^*\| \leqslant 1}} |w(z)| = \sup_{\substack{\|F: V_A \to V_B^*\| \leqslant 1 \\ \mathrm{rk}(F) \leqslant L_{\leftrightarrow} \\ \mathrm{rk}(F) \leqslant L_{\leftrightarrow}}} \mathrm{tr}\left[\widetilde{z}_G \ F^*\right].$$

As follows from elementary linear algebra, an operator $F : V_A \to V_B^*$ satisfies $\operatorname{rk}(F) \leq L_{\leftrightarrow}$ and $||F|| \leq 1$ if and only if it can be factorised as $F = f_2 f_1$, where $f_1 : V_A \to W$, $f_2 : W \to V_B^*$ are linear maps, W is a suitable Banach space of dimension dim $W \leq L_{\leftrightarrow}$, and $||f_1||, ||f_2|| \leq 1$. Using this observation, we can rewrite the upper bound in the above inequality as

$$\beta_{\leftrightarrow}(G) \leqslant \sup_{\substack{\dim W \leqslant L_{\leftrightarrow} \\ \|f_1: V_A \to W\|, \|f_2: W \to V_B^*\| \leqslant 1}} \operatorname{tr}\left[\widetilde{z}_G f_1^* f_2^*\right].$$
(A6)

Defining the tensors $u \in V_A^* \otimes W$ and $v \in V_B^* \otimes W^*$ such that $\tilde{u} = f_1$ and $\tilde{v} = f_2^*$, we can rewrite

$$\operatorname{tr}\left[\widetilde{z}_G f_1^* f_2^*\right] = v\left((\widetilde{z}_G \otimes \operatorname{Id}_W)(u)\right).$$

At the same time, the constraints $||f_1 : V_A \to W||$, $||f_2 : W \to V_B^*|| \le 1$ become simply $||u||_{V_A^* \otimes_{\varepsilon} W}$, $||v||_{V_B^* \otimes_{\varepsilon} W^*} \le 1$. Using once again (44), the bound in (A6) translates to

$$\begin{split} \beta_{\leftrightarrow}(G) &\leqslant \sup_{\substack{\dim W \leqslant L_{\leftrightarrow} \\ \|u\|_{V_{A}^{*} \otimes_{\varepsilon} W}, \ \|v\|_{V_{B}^{*} \otimes_{\varepsilon} W^{*}} \leqslant 1}} v\left((\widetilde{z}_{G} \otimes \operatorname{Id}_{W})(u)\right) \\ &= \sup_{\substack{\dim W \leqslant L_{\leftrightarrow} \\ \|u\|_{V_{A}^{*} \otimes_{\varepsilon} W} \leqslant 1}} \|(\widetilde{z}_{G} \otimes \operatorname{Id}_{W})(u)\|_{V_{B} \otimes_{\pi} W} \\ &= \sup_{\substack{\dim W \leqslant L_{\leftrightarrow} \\ \dim W \leqslant L_{\leftrightarrow}}} \|\widetilde{z}_{G} \otimes \operatorname{Id}_{W} : V_{A}^{*} \otimes_{\varepsilon} W \to V_{B} \otimes_{\pi} W \| \, . \end{split}$$

which proves the first upper bound in (A2). To obtain the other inequalities, we write

$$\begin{split} \|\widetilde{z}_{G} \otimes \operatorname{Id}_{W} : V_{A}^{*} \otimes_{\varepsilon} W \to V_{B} \otimes_{\pi} W \| \stackrel{1}{\leqslant} L_{\leftrightarrow} \|\widetilde{z}_{G} \otimes \operatorname{Id}_{W} : V_{A}^{*} \otimes_{\pi} W \to V_{B} \otimes_{\pi} W \| \\ \stackrel{2}{=} L_{\leftrightarrow} \|\widetilde{z}_{G} : V_{A}^{*} \to V_{B} \| \\ \stackrel{3}{=} L_{\leftrightarrow} \|z_{G}\|_{V_{A} \otimes_{\varepsilon} V_{B}} . \end{split}$$

The above steps are easy to justify: 1: we employed the inequality

$$\|\cdot\|_{V^*_A\otimes_{\varepsilon}W} \ge \frac{1}{\dim W} \|\cdot\|_{V^*_A\otimes_{\pi}W} \ge \frac{1}{L_{\leftrightarrow}} \|\cdot\|_{V^*_A\otimes_{\pi}W},$$

which derives from (18) (in turn proven in [32, Proposition 21]); 2: follows because the extreme points of the unit ball of $V_A^* \otimes_{\pi} W$ are product vectors; 3: is an application of (45). \Box

Remark. By the same kind of arguments, one can also show that sharing a physical system of bounded dimension does not help to increase the bias by more than a constant factor. We omit the details.

Finally, let us emphasise that here we have shown an upper bound for the bias of an XOR game with back-and-forth communication. In the work [58], the authors studied classical XOR games with both one-way classical communication and one-way quantum communication. It turns out that in that case, the bias of the the games can be exactly expressed in terms of certain norms of the corresponding operator $\tilde{z}_G : \ell_{\infty} \to \ell_1$.

Appendix B. Every Normed Space is 2-Isomorphic to a Base Norm Space

In this appendix we justify our choice of characterising the intrinsic difference between global and local strategies in XOR games by means of the projective/injective ratio as defined by (16) instead of (27), as discussed in Sect. 1.3. This corresponds to letting the optimisation run over all pairs of Banach spaces of fixed dimensions instead of restricting it to the base norm spaces alone, and does not lead to a significant loss of information because of the inequalities (29) and (30), whose proof we present here. Let us start with a preliminary result.

Lemma B.1. Every Banach space (possibly infinite-dimensional) is 2-isomorphic to a base norm space.

Proof. Let *X* be a Banach space. Pick a unit vector $x \in X$ such that ||x|| = 1, and consider the associated norming functional $x^* \in X^*$, which satisfies $||x^*|| = 1$ and $x^*(x) = 1$. Calling B_X the unit ball of *X*, construct the set $F := (x^*)^{-1}(1/2) \cap B_X$, and then set $B := \operatorname{cl} \operatorname{conv} (F \cup (-F))$, where the closure is possibly needed only in the infinite-dimensional case. It is not difficult to verify that *B* is the unit ball of the base norm space induced on *X* by the positive cone $\mathbb{R}_+ \cdot F = \{x \in X : x^*(x) \ge ||x||/2\}$ and the unit functional $u := 2x^*$ [2, p. 26]. Since $B \subseteq B_X$, it suffices to check that $B_X \subseteq 2B$ to establish the claim. To this end, we pick $y \in B_X$ and we check that $\frac{y}{2} \in B$. We can assume without loss of generality that $x^*(y) \ge 0$, while $|x^*(y)| \le 1$ holds by construction. We now distinguish two cases.

• If $x^*(y) \ge 1/2$, we can write

$$\frac{y}{2} = x^*(y) \frac{y}{2x^*(y)} \in x^*(y) B \subseteq B$$
,

where we used the fact that $2x^*(y) \ge 1$.

• The case where $0 \le x^*(y) < 1/2$ is significantly less transparent. Figure 2 conveys the geometric intuition behind the proof. An analytical argument is as follows. Call $k := x^*(y)$, and define the two vectors

$$z_{\pm} := \frac{1}{2(1 \mp k)} \left(\pm (1 \mp 2k)x + y \right).$$

Observe that z_+ lies at the intersection of the segment joining y and x with the plane $x^* = 1/2$. Analogously, z_- lies at the intersection of the segment joining y and -x with the plane $x^* = -1/2$. In particular, $z_{\pm} \in B$. We now try to obtain a multiple of y by taking a convex combination of z_+ and z_- . Setting

$$p(k) := \frac{(1+2k)(1-k)}{2(1-2k^2)}$$

which satisfies $1/2 \leq p(k) < 1$ for all $0 \leq k < 1/2$, we can write

$$\frac{y}{2(1-2k^2)} = p(k)z_+ + (1-p(k))z_- \in B.$$

By rescaling the vector on the l.h.s. we see that $y/2 \in B$.

This concludes the proof. \Box

We are now ready to prove the inequalities (29) and (30) discussed in the main text.

Lemma B.2. *The functions* r(n, m) *and* $r_{bn}(n, m)$ *defined by* (16) *and* (27) *satisfy* (29) *and* (30):

$$r_{\rm bn}(n,m) \leq 4r(n,m),$$

 $r_{\rm bn}(n,m) \leq 2 + r(n-1,m-1),$

for all integers $n, m \ge 2$.

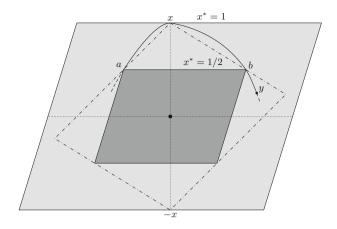


Fig. 2. Geometric intuition behind the proof of Lemma B.2, to which we refer for notation. The plane depicted is that spanned by the vectors *x* and *y*. From now on, all sets we consider are understood to be intersected with this plane. The curved solid and dashed line represents part of the boundary of the unit ball B_X (we set ||y|| = 1 for simplicity). The points *a* and *b* are the intersections of the line $x^* = 1/2$ with the boundary of B_X . The associated parallelogram *B* is depicted in dark grey colour, while 2*B* is in light grey colour. Then, an informal version of the argument is as follows. Since between *x* and *b* the boundary of B_X lies above the dash-dotted line, convexity implies that it must lie below that same line between *b* and *y*. This ensures that *y* belongs to the dash-dotted tilted parallelogram, which is readily verified to fit inside 2*B*

Proof. Lemma B.1 proves that for all Banach spaces X there is a base norm Banach space X' such that $d(X, X') \leq 2$, where d is the Banach–Mazur distance (49). We apply this to a pair of finite-dimensional Banach spaces X, Y, with dim X = n and dim Y = m, obtaining two base norm Banach spaces X' and Y' of the same dimension that are 2-isomorphic to X and Y, respectively. We find that

$$\begin{aligned} r_{\mathrm{bn}}(n,m) &\leqslant \rho(X',Y') \\ &\leqslant \mathrm{d}(X',X)\rho(X,Y') \\ &\leqslant \mathrm{d}(X',X)\,\mathrm{d}(Y',Y)\rho(X,Y) \\ &\leqslant 4\rho(X,Y)\,, \end{aligned}$$

where we used (65) twice, once for each of the two arguments of ρ (this is possible as ρ is symmetric, see (61)). Taking the infimum over all pairs *X*, *Y* yields (29).

We now move on to proving the second inequality (30). The main idea of the argument is to construct, given a pair of Banach spaces X, Y of dimensions n - 1, m - 1, another pair of base norm spaces X', Y' of dimensions n, m such that $\rho(X', Y') \approx \rho(X, Y)$. This can be done by setting $X' := X \oplus_{\infty} \mathbf{R}$, where $||(x, a)||_{X'} = \max\{||x||_X, |a|\}$ for all $x \in X$ and $a \in \mathbf{R}$, and analogously for Y'. It is not difficult to check that $|| \cdot ||_{X'}$ is in fact the base norm induced by the cone $C := \{(x, a) : a \ge ||x||_X\}$ and the order unit $u_{X'} \in (X')^*$ defined by $u_{X'}(x, a) = a$ for all $x \in X$ and $a \in \mathbf{R}$. Thus, X' and Y' are base norm spaces. Incidentally, this is a systematic way of associating 'centrally symmetric' GPTs to Banach spaces, see [32, Section 6.1]. We now proceed to show that

$$\rho(X', Y') \leqslant 2 + \rho(X, Y), \tag{B1}$$

using a similar technique to that employed in the proof of [32, Proposition 26]. Take

$$z' = \begin{pmatrix} z & s \\ t & a \end{pmatrix} \in X' \otimes Y' \,,$$

where $z \in X \otimes Y$, $s \in X$, $t \in Y$, and $a \in \mathbf{R}$. Using the fact that a unit functional $\varphi \in B_{(X')^*}$ acts as $\varphi(x, a) = px^*(x) \pm (1 - p)a$, for some $x^* \in B_{X^*}$ and $p \in [0, 1]$, it is not difficult to show that

$$\|z'\|_{X'\otimes_{\varepsilon}Y'} = \max\left\{\|z\|_{X\otimes_{\varepsilon}Y}, \|s\|, \|t\|, |a|\right\}.$$
 (B2)

We now give an upper estimate of the corresponding projective norm. Taking vectors $x_i \in X$ and $y_i \in Y$ such that $z = \sum_i x_i \otimes y_i$ and $||z||_{X \otimes_{\pi} Y} = \sum_i ||x_i|| ||y_i||$, and for an arbitrary $p \in [0, 1]$, we consider the decomposition

$$z' = \begin{pmatrix} 0 & s \\ 0 & pa \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ t & (1-p)a \end{pmatrix} + \begin{pmatrix} z & 0 \\ 0 & 0 \end{pmatrix}$$

= $(s, pa) \otimes (0, 1) + (0, 1) \otimes (t, (1-p)a) + \sum_{i} (x_i, 0) \otimes (y_i, 0),$

which yields the estimate

$$\begin{aligned} \|z'\|_{X'\otimes_{\pi}Y'} &\leq \min_{p\in[0,1]} \left\{ \max\{\|s\|, p|a|\} + \max\{\|t\|, (1-p)|a|\} + \sum_{i} \|x_{i}\| \|y_{i}\| \right\} \\ &= \max\{\|s\| + \|t\|, |a|\} + \sum_{i} \|x_{i}\| \|y_{i}\| \\ &= \max\{\|s\| + \|t\|, |a|\} + \|z\|_{X\otimes_{\pi}Y} \\ &\leq \max\{\|s\| + \|t\|, |a|\} + \rho(X, Y) \|z\|_{X\otimes_{\varepsilon}Y} \\ &\leq (2 + \rho(X, Y)) \|z'\|_{X'\otimes_{\varepsilon}Y'}. \end{aligned}$$
(B3)

Optimising over all $z' \in X' \otimes Y'$ and using (15) gives the estimate in (B1). We can now write

$$r_{bn}(n,m) \leq \inf_{\substack{\dim X = n-1 \\ \dim Y = m-1}} \rho(X',Y')$$
$$\leq \inf_{\substack{\dim X = n-1 \\ \dim Y = m-1}} \{2 + \rho(X,Y)\}$$
$$= 2 + r(n-1,m-1).$$

This concludes the proof. \Box

Appendix C. Functional Analytic Tools

1. The ℓ -norm and the MM^* -estimate. Let X be a real Banach space. Given a linear map $T : \ell_2^n \to X$, the ℓ -norm of T is defined as

$$\ell_X(T) = \mathbf{E} \left\| \sum_{i=1}^n g_i T |e_i\rangle \right\|_X,\tag{C1}$$

where $(|e_i\rangle)_{i=1}^n$ is an orthormal basis of \mathbb{R}^n and $(g_i)_{i=1}^n$ is a sequence of independent N(0, 1) Gaussian random variables. We point out that several authors prefer to define ℓ -norms via the second moment, i.e. $\ell_X(T) = \left(\mathbf{E} \left\|\sum_{i=1}^n g_i T |e_i\rangle\right\|_X^2\right)^{1/2}$. However both definitions give equivalent norms in view of Theorem 2. Also, note that the invariance of the Gaussian measure under unitary transformations implies that $\ell_X(T) = \ell_X(T \circ U)$ for every unitary $U : \ell_2^n \to \ell_2^n$ and, in particular, (C1) does not depend on the choice of orthonormal basis.

The following theorem will be crucial for us.

Theorem C.1 (*MM*^{*}-estimate). Let X be an n-dimensional Banach space. Then there exists an isomorphism $T : \ell_2^n \to X$ such that

$$\ell_X(T)\ell_{X^*}((T^{-1})^*) \leqslant Cn\log n.$$

That statement is a direct consequence of Lewis' theorem ([59, Theorem 3.1]) and a well known estimate on the so-called *K*-convexity constant of a Banach space. The reader can find a detailed proof of Theorem C.1 in [59, Theorem 3.11] or [60, Theorem 7.10].

2. *Some Gaussian inequalities.* We will make use of Khintchine–Kahane inequalities (see for instance [59, Corollary 4.9] or, for optimal constants, [61, Corollary 3]).

Theorem 2 (Khintchine–Kahane inequalities). For every $1 there exists a universal constant <math>C_p > 0$ such that for every Banach space X and every sequence of elements $(x_i)_{i=1}^n \subset X$ we have

$$\mathbf{E} \left\| \sum_{i=1}^{n} g_{i} x_{i} \right\|_{X} \leq \left(\mathbf{E} \left\| \sum_{i=1}^{n} g_{i} x_{i} \right\|_{X}^{p} \right)^{\frac{1}{p}} \leq C_{p} \mathbf{E} \left\| \sum_{i=1}^{n} g_{i} x_{i} \right\|_{X}$$

We will also make use of Chevet's inequality ([53, Theorem 43.1]).

Theorem 3 (Chevet's inequality). Let X and Y be real Banach spaces. Define the Gaussian random tensor $z = \sum_{i=1}^{m} \sum_{j=1}^{n} g_{ij} x_i \otimes y_j \in X \otimes Y$, where (g_{ij}) are independent N(0, 1) Gaussian random variables, and $(x_i)_{i=1}^m \subset X$, $(y_j)_{j=1}^n \subset Y$ are sequences of elements. Then,

$$\mathbf{E} \| z \|_{X \otimes_{\varepsilon} Y} \leq \sup_{x^* \in B_{X^*}} \left(\sum_{i=1}^m |x^*(x_i)|^2 \right)^{\frac{1}{2}} \mathbf{E} \left\| \sum_{j=1}^n g_j y_j \right\|_{Y} + \sup_{y^* \in B_{Y^*}} \left(\sum_{j=1}^n |y^*(y_j)|^2 \right)^{\frac{1}{2}} \mathbf{E} \left\| \sum_{i=1}^m g_i x_i \right\|_{X},$$

where $(g_i)_i$ is a sequence of independent N(0, 1) Gaussian random variables. Note that, given a Banach space Z and $(z_i)_{i=1}^n \subset Z$, we have

$$\sup_{z^* \in B_{Z^*}} \left(\sum_{i=1}^n |z^*(z_i)|^2 \right)^{\frac{1}{2}} = \|T : \ell_2^n \to Z\|.$$

Here *T* is the linear map defined by $T|e_i\rangle = z_i$ for every $i = 1, \dots, n$, where $(|e_i\rangle)_i$ is an orthonormal basis of ℓ_2^n .

Here is a typical application of Chevet's inequality. Fix integers m, n, and consider $(x_i)_{1 \le i \le m^2}$ and $(y_j)_{1 \le j \le n^2}$ orthonormal bases of M_m^{sa} and M_n^{sa} respectively, with respect to the Hilbert–Schmidt inner product. We form the random tensor

$$z = \sum_{i=1}^{m^2} \sum_{j=1}^{n^2} g_{ij} x_i \otimes y_j,$$
 (C2)

where (g_{ij}) are independent N(0, 1) Gaussian random variables.

Corollary C.4. Let z be defined as in (C2). Remember that we denote by $S_p^{n,sa}$ the space of $n \times n$ Hermitian matrices equipped with the Schatten norm $\|\cdot\|_p$. Then

$$\mathbf{E} \| z \|_{S_1^{m, \mathrm{sa}} \otimes_{\varepsilon} S_1^{n, \mathrm{sa}}} \leq C \sqrt{mn} \max\{m, n\},$$

$$\mathbf{E} \| z \|_{S_\infty^{m, \mathrm{sa}} \otimes_{\varepsilon} S_\infty^{m, \mathrm{sa}}} \leq C \max\{\sqrt{m}, \sqrt{n}\}.$$

Proof. In both cases we apply Theorem 3 and need to estimate all quantities appearing on the r.h.s. The random matrix $G_m = \sum_{j=1}^m g_j x_j$ is distributed according to the Gaussian Unitary Ensemble. It is well known (see for example [60, Proposition 6.24]) that as *m* tends to infinity,

$$\mathbf{E} \|G_m\|_{\infty} \sim 2\sqrt{m}, \quad \mathbf{E} \|G_m\|_1 \leqslant m^{3/2}.$$

On the other hand, it is also well known (and easy to check) that

$$||\mathrm{Id}: S_2^{m,\mathrm{sa}} \to S_\infty^{m,\mathrm{sa}}|| \leqslant 1, \quad ||\mathrm{Id}: S_2^{m,\mathrm{sa}} \to S_1^{m,\mathrm{sa}}|| \leqslant \sqrt{m}.$$

Since $S_2^{m,sa}$ is isometric to $\ell_2^{m^2}$, we can apply Theorem 3 to conclude that

$$\mathbf{E} \| z \|_{S_1^{m, \mathrm{sa}} \otimes_{\varepsilon} S_1^{n, \mathrm{sa}}} \leqslant n^{\frac{3}{2}} \sqrt{m} + m^{\frac{3}{2}} \sqrt{n} \leqslant 2\sqrt{mn} \max\{m, n\},$$
$$\mathbf{E} \| z \|_{S_{\infty}^{m, \mathrm{sa}} \otimes_{\varepsilon} S_{\infty}^{n, \mathrm{sa}}} \leqslant 2\sqrt{m} + 2\sqrt{n} \leqslant 4 \max\{m, n\};$$

hence the result follows. \Box

Finally, we will use the following lemma, whose proof is elementary.

Lemma C.5 (Contraction principle). Let $(\alpha_i)_{i=1}^n$ and $(\beta_i)_{i=1}^n$ be two sequences of numbers with $0 \leq \alpha_i \leq \beta_i$ for every *i*. Let $(g_i)_{i=1}^n$ be a sequence of independent N(0, 1) Gaussian random variables. Then, for every Banach space X and every $x_1, \dots, x_n \in X$, we have

$$\mathbf{E}\left\|\sum_{i=1}^{n}\alpha_{i}g_{i}x_{i}\right\|_{X}\leq\mathbf{E}\left\|\sum_{i=1}^{n}\beta_{i}g_{i}x_{i}\right\|_{X}.$$

3. Dvoretkzy–Milman theorem. We also need Milman's version of Dvoretzky theorem (see e.g. [60, Theorem 7.19]). Let $\|\cdot\|_X$ be a norm on \mathbb{R}^n and consider the space $X = (\mathbb{R}^n, \|\cdot\|_X)$. The **Dvoretzky dimension** of X is defined as

$$k_*(X) = \left(\frac{\ell_X(\mathrm{Id})}{\|\mathrm{Id}:\ell_2^n \to X\|}\right)^2.$$

Theorem C.6 (Dvoretzky–Milman theorem). *Consider a normed space* $X = (\mathbf{R}^n, \| \cdot \|_X)$ and let $E \subseteq \mathbf{R}^n$ be a random subspace of dimension $k \leq c k_*(X)$. Then, with large probability,

$$\frac{\ell_X(\mathrm{Id})}{2\sqrt{n}}|x| \leqslant \|x\|_X \leqslant \frac{2\ell_X(\mathrm{Id})}{\sqrt{n}}|x|$$

for every $x \in E$, where $|\cdot|$ is the standard Euclidean norm on \mathbb{R}^n . In particular, the space $X \cap E$ is 4-Euclidean.

Remark. In Theorem C.6 it is understood that E is distributed according to the Haar measure on the Grassmann manifold (see e.g. [60]). The expression 'with large probability' means that the probability of failure tends to zero exponentially fast as n tends to infinity; we need only to know that the intersection of two such events of large probability is nonempty.

4. John ellipsoid. The following theorem is a classical result about convex bodies (see [62] for a modern proof).

Theorem C.7 (John's theorem). For every n-dimensional normed space X with unit ball B_X , there is a unique ellipsoid \mathcal{E} of maximal volume under the constraint $\mathcal{E} \subseteq B_X$. The ellipsoid \mathcal{E} is called the John ellipsoid of X and satisfies $B_X \subseteq \sqrt{n\mathcal{E}}$. Consequently, we have $d(X, \ell_2^n) \leq \sqrt{n}$.

We also use a variant of John's theorem. It can be for example deduced from [59, Corollary 3.9].

Proposition C.8. Let X be a finite-dimensional normed space with dim $(X) \ge n$. Then there exist maps $u : \ell_2^n \to X$ and $v : X \to \ell_2^n$ such that $vu = \mathrm{Id}_{\ell_2^n}$, ||u|| = 1 and $||v|| \le \sqrt{n}$.

We will rely on a technical result which guarantees that certain normed spaces contains large-dimensional cubes; in that formulation it is due to Vershynin [63] (improving on Rudelson [7]).

Theorem C.9 (Theorem 6.2 in [63]). Let $X = (\mathbb{R}^n, \|\cdot\|_X)$ be a *n*-dimensional normed space whose John ellipsoid is B_2^n . Let *P* be an orthogonal projection and $k = \operatorname{rank}(P)$. Then there are $m \ge ck/\sqrt{n}$ contact points $(x_j)_{1 \le j \le m}$ such that

$$\max_{1 \leq j \leq m} |\langle x, x_j \rangle| \leq ||x||_X \leq C \sqrt{\frac{n}{k}} \ell_X(P) \max_{1 \leq j \leq m} |\langle x, x_j \rangle|$$
(C3)

for every $x \in \text{span}\{Px_j : 1 \leq j \leq m\}$. In particular, the space X contains a subspace which is R-isomorphic to ℓ_{∞}^m for $R = C\ell_X(P)\sqrt{n/k}$.

5. *p*-summing norms. Let $u : X \to Y$ be a linear map between finite-dimensional normed spaces. Fix $p \in [1, \infty)$; we only need p = 1 and p = 2 in the present paper. For an integer N, we define a quantity $\pi_p^{(N)}(u)$ to be the smallest constant K such that, for every N vectors $x_1, \dots, x_N \in X$, we have

$$\left(\sum_{k=1}^{N} \|u(x_k)\|^p\right)^{1/p} \leqslant K \sup_{\phi \in B_{Y^*}} \left(\sum_{k=1}^{N} |\phi(x_k)|^p\right)^{1/p}.$$
 (C4)

The quantity $\pi_p(u) = \sup\{\pi_p^{(N)}(u) : N \ge 1\}$ is called the *p*-summing norm of the operator *u*.

Proposition C.10. Consider finite-dimensional normed spaces X, Y, and a linear operator $u : X \to Y$. We have

1. $\pi_1^{(N)}(u) \ge \pi_2^{(N)}(u),$ 2. $\pi_2^{(\dim X)}(u) \ge \frac{1}{\sqrt{2}}\pi_2(u),$ 3. $\pi_2(\operatorname{Id}_X) = \sqrt{\dim X}.$

A general reference about *p*-summing norms (with detailed bibliography) is [53]; parts 1-3 of Proposition C.10 appear there respectively as Proposition 9.6, Theorem 18.4 and Proposition 9.11.

We also need a specific result about the 1-summing norm of the identity map on ℓ_1^n , which appears as [64, Theorem 2(4)]: we have

$$\pi_1(\mathrm{Id}_{\ell_1^n}) = \frac{n}{\mathbf{E}\left|\sum_{i=1}^n \varepsilon_i\right|},\tag{C5}$$

where (ε_i) is a sequence of independent random variables with $\mathbf{P}(\varepsilon_i = 1) = \mathbf{P}(\varepsilon_1 = -1) = \frac{1}{2}$. For a more transparent derivation, one may use the fact that ℓ_1^n has enough symmetries in the sense of [53, §16], which implies that $\pi_1(\mathrm{Id}_{\ell_1^n}) = n/f(\ell_1^n, \ell_\infty)$ (see e.g. [65]). In turn, the quantity $f(X, \ell_\infty)$ (defined in (50) and also referred to as the **projection constant** of a normed space *X*, see [53, §32]) can be calculated directly when $X = \ell_1^n$ and equals $\mathbf{E} |\varepsilon_1 + \cdots + \varepsilon_n|$; an early reference for the last result is [66, Theorem 3].

6. Non-commutative Grothendieck inequality. Let us recall Grothendieck's inequality for bilinear forms on C^* -algebras. Here, we state [67, Theorem 1.1], which improved the original proof in [68].

Theorem C.11 (Grothendieck's inequality for C^* -algebras). Let $V : \mathcal{A} \times \mathcal{B} \to \mathbb{C}$ be a bilinear form on a pair of C^* -algebras \mathcal{A} and \mathcal{B} . Then, there exist two states φ_1 and φ_2 on \mathcal{A} and two states ψ_1 and ψ_2 on \mathcal{B} such that

$$|V(x, y)| \leq \|V\| \left(\varphi_1(x^*x) + \varphi_2(xx^*)\right)^{\frac{1}{2}} \left(\psi_1(y^*y) + \psi_2(yy^*)\right)^{\frac{1}{2}} \quad \forall x \in \mathcal{A}, \ y \in \mathcal{B},$$
(C6)

where

$$\|V\| := \sup_{\substack{x \in \mathcal{A}, \ y \in \mathcal{B} \\ \|x\|, \|y\| \leqslant 1}} |V(x, y)|$$
(C7)

is the norm of V.

By applying the previous theorem to the particular case $\mathcal{A} = S_{\infty}^{n}$ (the *C**-algebra of $n \times n$ complex matrices endowed with the operator norm) and $\mathcal{B} = S_{\infty}^{m}$, we deduce the following corollary.

Corollary C.12. Let $z \in S_1^{n,sa} \otimes_{\varepsilon} S_1^{m,sa}$ be a tensor, and let $\tilde{z} : S_{\infty}^{n,sa} \to S_1^{m,sa}$ be the linear map associated with it according to (42). Then, there exists a state φ on S_{∞}^n such that

$$\|\widetilde{z}(x)\|_{S_1^{m,\mathrm{sa}}} \leqslant 2\sqrt{2} \|z\|_{S_1^{n,\mathrm{sa}} \otimes_{\varepsilon} S_1^{m,\mathrm{sa}}} (\varphi(x^2))^{\frac{1}{2}} \quad \forall x \in S_{\infty}^{n,\mathrm{sa}}.$$
 (C8)

Proof. We start by remarking that since $S_1^{k,sa}$ can be thought of as a (real) subspace of the (complex) Banach space of all $k \times k$ complex matrices endowed with the trace norm, denoted by S_1^k , we can consider z also as a tensor in $S_1^n \otimes_{\varepsilon} S_1^m$. According to [47, Claim 4.7], we have that

$$\|z\|_{S_1^n\otimes_{\varepsilon}S_1^m} \leqslant \sqrt{2} \|z\|_{S_1^{n,\mathrm{sa}}\otimes_{\varepsilon}S_1^{m,\mathrm{sa}}}$$

Indeed, to see this just notice that [47, Definition 4.3] and [47, Definition 4.6] correspond to $||z||_{S_1^{n,sa} \otimes_{\varepsilon} S_1^{m,sa}}$ and $||z||_{S_1^n \otimes_{\varepsilon} S_1^m}$, respectively. Now we consider the bilinear form V_z : $S_{\infty}^n \times S_{\infty}^m \to \mathbf{C}$ defined by $V_z(x, y) := \text{tr}[(x \otimes y)z]$, whose norm can be verified to coincide with the injective norm of the tensor *z*, i.e.

$$\|V_z\| = \|z\|_{S_1^n \otimes_{\varepsilon} S_1^m}.$$

Applying Theorem C.11 to V_z then yields

$$\begin{aligned} |\operatorname{tr}[\tilde{z}(x)y]| &= |\operatorname{tr}[(x \otimes y)z]| \\ &\leq ||z||_{S_1^n \otimes_{\varepsilon} S_1^m} \left(\varphi_1(x^*x) + \varphi_2(xx^*)\right)^{\frac{1}{2}} \left(\psi_1(y^*y) + \psi_2(yy^*)\right)^{\frac{1}{2}} \\ &\leq \sqrt{2} ||z||_{S_1^{n,\operatorname{sa}} \otimes_{\varepsilon} S_1^{m,\operatorname{sa}}} \left(\varphi_1(x^*x) + \varphi_2(xx^*)\right)^{\frac{1}{2}} \left(\psi_1(y^*y) + \psi_2(yy^*)\right)^{\frac{1}{2}} \end{aligned}$$

Taking the supremum over all $y \in S_{\infty}^{m, \text{sa}}$ such that $||y||_{\infty} \leq 1$, using the fact that $\psi(y^*y) \leq 1$ and $\psi(yy^*) \leq 1$ for all such y and for all states ψ on S_{∞}^m , and finally defining $\varphi := (\varphi_1 + \varphi_2)/2$, we obtain precisely (C8). \Box

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Communicated by M. M. Wolf.