Moments of Gaussian chaoses in Banach spaces*

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Abstract

We derive moment and tail estimates for Gaussian chaoses of arbitrary order with values in Banach spaces. We formulate a conjecture regarding two-sided estimates and show that it holds in a certain class of Banach spaces including $L_q$ spaces. As a corollary we obtain two-sided bounds for moments of chaoses with values in $L_q$ spaces based on exponential random variables.

Keywords: tail and moment inequalities; polynomials in independent random variables; Gaussian chaoses; Gaussian processes; metric entropy.

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1 Introduction

Multivariate polynomials in Gaussian variables have been extensively studied at least since the work of Wiener in the 1930s. They have found numerous applications in the theory of stochastic integration and Malliavin calculus [12, 22, 23], functional analysis [11], limit theory for $U$-statistics [9] or long-range dependent processes [29], random graph theory [12], and more recently computer science [7, 14, 19, 24]. While early results considered mostly polynomials with real coefficients, their vector-valued counterparts also appear naturally, e.g., in the context of stochastic integration in Banach spaces [20], in the study of weak limits of U-processes [9], as tools in characterization of various geometric properties of Banach spaces [11, 25, 26] or in the analysis of empirical covariance operators [1, 30]. Apart from applications, the theory of Gaussian polynomials has been studied for its rich intrinsic structure, with interesting interplay of analytic, probabilistic, algebraic and combinatorial phenomena, leading to many challenging problems. For a comprehensive presentation of diverse aspects of the theory we refer to the monographs [9, 11, 12, 17].

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An important aspect of the study of Gaussian polynomials is the order of their tail decay and growth of moments. In the real valued case the first estimates concerning this question, related to the hypercontractivity of the Ornstein-Uhlenbeck semigroup, were obtained by Nelson [21]. For homogeneous tetrahedral (i.e., affine in each variable) forms of arbitrary fixed degree two-sided estimates on the tails and moments were obtained in [15] (in particular generalizing the well-known Hanson-Wright inequality for quadratic forms). In [4] it was shown that the results of [15] in fact allow to obtain such estimates for all polynomials of degree bounded from above. Two-sided estimates for polynomials with values in a Banach space have been obtained independently by Borell [6], Ledoux [16], Arcones-Giné [5]. They are expressed in terms of suprema of certain empirical processes (see formula (1.5) below), which in general may be difficult to estimate (even in the real valued case).

In a recent paper [2] we considered Gaussian quadratic forms with coefficients in a Banach space and obtained upper bounds on their tails and moments, expressed in terms of quantities which are easier to deal with. In the real valued case our estimates reduce to the Hanson-Wright inequality, and for a large class of Banach-spaces (related to Pisier’s Gaussian property $\alpha$ and containing all type 2 spaces) they may be reversed. In particular for $L_q$ spaces with $1 \leq q < \infty$ they yield two-sided estimates expressed in terms of deterministic quantities. In the present work we generalize these estimates to polynomials of arbitrary degree.

Before presenting our main theorems (which requires an introduction of a rather involved notation) let us describe the setting and discuss in more detail some of the results mentioned above.

To this aim consider a Banach space $(F, \| \cdot \|)$. A (homogeneous, tetrahedral) $F$-valued Gaussian chaos of order $d$ is a random variable defined as

$$S = \sum_{1 \leq i_1 < i_2 < \cdots < i_d \leq n} a_{i_1 \ldots i_d} g_{i_1} \cdots g_{i_d}, \tag{1.1}$$

where $a_{i_1 \ldots i_d} \in F$ and $g_1, \ldots, g_n$ are i.i.d. standard Gaussian variables. As explained above the goal of this paper is to derive estimates on moments (defined as $\| S \|_p := (E \| S \|^p)^{1/p}$) and tails of $S$, more precisely to establish upper bounds which for some classes of Banach spaces, including $L_q$ spaces, can be reversed (up to constants depending only on $d$ and the Banach space, but not on $n$ or $a_{i_1 \ldots i_d}$). We restrict to random variables of the form (1.1), however it turns out that estimates on their moments will in fact allow to deduce moment and tail bounds for arbitrary polynomials in Gaussian random variables as well as for homogeneous tetrahedral polynomials in i.i.d. symmetric exponential random variables. In the sequel we will focus on decoupled chaoses

$$S' = \sum_{i_1, \ldots, i_d=1}^n a_{i_1 \ldots i_d}^{(1)} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)}, \tag{1.2}$$

where $(g_{i_k}^{(k)})_{i,k \geq 1}$ are independent $N(0,1)$ random variables – under natural symmetry assumptions, moments and tails of $S, S'$ are comparable up to constants depending only on $d$ (cf. Theorem A.9 in the Appendix). Moreover, as will we see in Proposition 2.11, estimating moments of general polynomials in i.i.d. standard Gaussian variables can be reduced to estimating moments of variables of the form (1.2).

For $d = 1$ and any $p \geq 1$ one has the following well-known estimate (cf. Lemma A.5)

$$\left\| \sum_{i=1}^n a_i g_i \right\|_p = \left( E \left[ \sum_{i=1}^n a_i g_i \right]^p \right)^{1/p} \sim E \left[ \sum_{i=1}^n a_i g_i \right] + \sqrt{p} \sup_{x \in B_2^n} \left\| \sum_{i=1}^n a_i x_i \right\|, \tag{1.3}$$

where, $B_2^n$ is the unit standard Euclidean ball in $\mathbb{R}^n$ and $\sim$ stands for a comparison up to universal multiplicative constants.
An iteration of the above inequality yields for chaoses of order 2,
\[
\left\| \sum_{i,j=1}^{n} a_{ij} g_i g_j \right\|_p \sim E \left\| \sum_{i,j=1}^{n} a_{ij} g_i g_j \right\| + \sqrt{p} E \sup_{x \in B_2^d} \left( \left\| \sum_{i,j=1}^{n} a_{ij} g_i x_j \right\| + \left\| \sum_{i,j=1}^{n} a_{ij} x_i g_j \right\| \right)
\]
\[+ p \sup_{x,y \in B_2^d} \left\| \sum_{i,j=1}^{n} a_{ij} x_i y_j \right\|, \tag{1.4}\]
where in the above formula and in the whole paper, \((g_j)_{j \geq 1}\) is an independent copy of \((g_i)_{i \geq 1}\).

For chaoses of higher order one gets an estimate
\[
\left\| \sum_{i_1, \ldots, i_d=1}^{n} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\|_p \sim^d \sum_{J \subseteq [d]} p^{d/2} E \sup_{x \in B_2^d} \left\| \sum_{i_1, \ldots, i_d=1}^{n} a_{i_1, \ldots, i_d} \prod_{j \in J} x_{i_j}^{(j)} \prod_{j \notin J} g_{i_j}^{(j)} \right\|, \tag{1.5}\]
where the supremum is taken over \(x^{(1)}, \ldots, x^{(n)}\) from the Euclidean unit ball and \(\sim^a\) stands for comparison up to constants depending only on the parameter \(a\). To the best of our knowledge the above inequality was for the first time established in [6] and subsequently reproved in various context by several authors [5, 16, 17].

The estimate (1.5) gives precise dependence on \(p\), but unfortunately is expressed in terms of expected suprema of certain stochastic processes, which are hard to estimate. In many situations this precludes effective applications. Let us note that even for \(d = 1\), the estimate (1.3) involves the expectation of a norm of a Gaussian random vector. Estimating such a quantity in general Banach spaces is a difficult task, which requires investigating the geometry of the unit ball of the dual of \(F\) (as described by the celebrated majorizing measure theorem due to Fernique and Talagrand). Therefore, in general one cannot hope to get rid of certain expectations in the estimates for moments.

Nevertheless, in some classes of Banach spaces (such as, e.g., Hilbert spaces, or more generally type 2 spaces) expectations of Gaussian chaoses can be easily estimated. The difficult part (also for \(d = 2\) and mentioned class of Banach spaces) is to estimate the terms in (1.4) and (1.5) which involve additional suprema over products of unit balls.

Even for \(d = 2\) and a Hilbert space, the term \(E \sup_{x \in B_2^2} \left\| \sum_{i,j=1}^{n} a_{ij} g_i x_j \right\|\) can be equivalently rewritten as the expected operator norm of a certain random matrix. Such quantities are known to be hard to estimate. Therefore, it is natural to seek inequalities which are expressed in terms of deterministic quantities and expectations of some \(F\)-valued polynomial chaoses, but do not involve expectations of additional suprema of such polynomials. This was the motivation behind the article [2], concerning the case \(d = 2\) and containing the following bound, valid for \(p \geq 1\) ([2, Theorem 4]),
\[
\left\| \sum_{i,j=1}^{n} a_{ij} g_i g_j \right\|_p \leq C \left( E \left\| \sum_{i,j=1}^{n} a_{ij} g_i g_j \right\| + E \left\| \sum_{i,j=1}^{n} a_{ij} g_i g_j \right\| \right)
\]
\[+ p^{1/2} \sup_{x, y \in B_2^2} E \left\| \sum_{i,j=1}^{n} a_{ij} g_i x_j \right\| + p^{1/2} \sup_{x, y \in B_2^2} E \left\| \sum_{i,j=1}^{n} a_{ij} x_i g_j \right\| + p \sup_{x, y \in B_2^2} \left\{ \sum_{i,j=1}^{n} a_{ij} x_i y_j \right\}, \tag{1.6}\]
where (with a slight abuse of notation) we denote \(B_2^{n^2} = \{(x_{ij})_{i,j=1}^{n^n} : \sum_{i,j=1}^{n} x_{ij}^2 \leq 1\} \).
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Let us point out that even though the inequalities we have presented so far as well as those we are about to discuss in the subsequent part of the article are formulated for general, possibly infinite dimensional, Banach spaces, the random variables involved take values in finite dimensional subspaces spanned by the coefficients of the polynomials in question. Therefore, as long as the constants in the inequalities are universal or do not depend on the particular subspace but just on some numerical characteristic of the space (e.g., the type constant), there is no loss in generality in assuming that the space $F$ is finite dimensional. In particular one can always assume without loss of generality that $F$ as a linear space equals $\mathbb{R}^m$ for some positive integer $m$. This phenomenon is well known in the local theory of Banach spaces.

It can be shown that in general inequality (1.6) cannot be reversed. However, it turns out to be two-sided in a certain class of Banach spaces containing $L_q$ spaces (see Section 2.1 below). This observation gives rise to the question of obtaining similar results for arbitrary $d$. Building on ideas and techniques developed in [15] we are able to give an answer to it. In our main result, Theorem 2.1, we provide an upper bound on moments of decoupled chaoses of order $d$, which generalizes (1.6). We also obtain lower bounds, which we conjecture to be in fact two-sided (see Conjecture 2.2), and in Section 2.1 we identify a large class of Banach spaces for which our upper and lower bounds do match.

Let us briefly comment on the proof of Theorem 2.1. The lower bound for moments relies on a rather straightforward reduction to the real-valued case, treated in [15]. The much more involved proof of the upper bound is based on an inductive approach. The inequality (1.6) serves as the base of induction, while (1.3) allows to reduce the induction step to an estimate of an expectation of a supremum of a certain canonical Gaussian process, which turns out to be the heart of the problem. In order to obtain such an estimate we apply a variant of the chaining method (see the monograph [28]), which requires bounds on the entropy numbers for the indexing set of the process in its intrinsic metric. In our case this metric is given via a norm on a tensor product of $F^*$ and several Euclidean spaces, whereas the indexing set is a Cartesian product of the corresponding unit balls. The bounds on entropy numbers are obtained by a variant of the volumetric argument as well as Sudakov and dual Sudakov minoration leading to expectations of suprema of other Gaussian processes, which can be estimated by using the induction hypotheses. This approach in a sense parallels the one used in [15] for the real-valued case, with (1.6) replacing the classical Hanson-Wright inequality, however it presents some additional difficulties related to the geometry of the unit ball in the space $F^*$. Let us also remark that this approach cannot be used to pass from $d = 1$, i.e., the inequality (1.3), to $d = 2$, i.e., the inequality (1.6) (see Remark 3.7). While the proof of (1.6) presented in [2] relies on a similar set of tools (chaining arguments, Gaussian concentration) some of the technical estimates of entropy numbers are obtained differently than in the induction step $d \rightarrow d + 1$ for $d \geq 2$.

The paper is organized as follows. In the next section we set up the notation and formulate the main results, in particular the pivotal bound for moments of homogeneous tetrahedral Gaussian chaoses in an arbitrary Banach space (Theorem 2.1). We also present its consequences: tail and moment estimates for arbitrary Gaussian polynomials, two-sided bounds in special classes of Banach spaces, inequalities for tetrahedral homogeneous forms in i.i.d. symmetric exponential variables. In Section 3, in Theorem 3.1, we formulate a key inequality for the supremum of a certain Gaussian processes and derive certain entropy bounds to be used in its proof, presented in Section 4. In Section 5 we use Theorem 3.1 to prove Theorem 2.1 from which we deduce all the remaining claims of Section 2. The Appendix contains certain basic facts concerning Gaussian processes and Gaussian polynomials used in the proofs. At the end of the article we provide a glossary explaining the notation.
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2 Notation and main results

In this section we introduce the most basic notation used in the article and formulate our main results. Since some additional notation will be introduced as the proofs develop, for the reader’s convenience at the end of the article we include a glossary of the most important symbols appearing in the text.

We write \([n]\) for the set \([1, \ldots, n]\). Throughout the article \(C\) (resp. \(C(\alpha)\)) will denote an absolute constant (resp. a constant which may depend on \(\alpha\)) which may differ at each occurrence. By \(A\) we typically denote a finite multi-indexed matrix \((a_{i_1 \ldots i_d})_{1 \leq i_1, \ldots, i_d \leq n}\) of order \(d\) with values in a Banach space \(\langle F, \|\cdot\|\rangle\). If \(I = (i_1, \ldots, i_d)\), \(i_1, \ldots, i_d \in [n]\) and \(I \subset [d]\), then we define \(i_I := (i_j)_{j \in I}\). To simplify the notation we will also often treat \(i_I\) as a stand-alone multi-index, with the meaning that each \(i_j, j \in I\) runs trough \([n]\).

We will also often suppress the range of summation. Unless stated otherwise the sums \(\sum\) when we write \(P\) of (unordered) partitions of \(B\), the space of (unordered) partitions of \([1, \ldots, n]\), then we define \(I, P \ni \) of \(P\), \(I\) of \(P\), \(I\) of \(P\).

We write \(\|\cdot\|\) as a stand-alone multi-index, with the meaning that each \(i\) ranges over \(I, P \ni \) of \(P\), \(I\) of \(P\).

In particular for \(d = 3\), \(I = [2, 3]\),

\[
\sum_{i_1, i_2, i_3 = 1}^{n} a_{i_1 i_2 i_3} x_{i_1 i_2 i_3}, \quad \sum_{i_j = 1}^{n} x_{i_j}^2 = \sum_{i_2, i_3 = 1}^{n} x_{i_2 i_3}^2.
\]

We note that all the multi-linear forms we consider are given by finite sums. Standard arguments allow to extend our inequalities to the case of infinite multiple series, but we do not pursue this direction.

In what follows we will often identify the space \(\mathbb{R}^{n}^{\otimes d}\) of \(d\)-indexed matrices with the space \(\mathbb{R}^{n^d}\). In particular \(B_2^{n^d}\) will stand for the unit Euclidean ball in \(\mathbb{R}^{n^d}\), i.e., \(B_2^{n^d} = \{ (b_{ij})_{ij \in [n]^d} \in \mathbb{R}^{n^d} : \sum b_{ij}^2 \leq 1 \}\).

If \(I\) is a finite set then \(|I|\) stands for its cardinality and by \(\mathcal{P}(I)\) we denote the family of (unordered) partitions of \(I\) into nonempty, pairwise disjoint sets. Note that if \(I = \emptyset\) then \(\mathcal{P}(I)\) consists only of the empty partition \(\emptyset\).

With a slight abuse of notation we write \((\mathcal{P}, \mathcal{P}') \in \mathcal{P}(I)\) if \(\mathcal{P} \cup \mathcal{P}' \in \mathcal{P}(I)\) and \(\mathcal{P} \cap \mathcal{P}' = \emptyset\).

Let \(\mathcal{P} = \{ I_1, \ldots, I_k \}\), \(\mathcal{P}' = \{ J_1, \ldots, J_m \}\) be such that \((\mathcal{P}, \mathcal{P}') \in \mathcal{P}(\{[d]\})\). Then we define

\[
\|A\|_{\mathcal{P}', \mathcal{P}} := \sup \left\{ \mathbb{E} \left\| \sum_{i_{1}, \ldots, i_{d}} a_{i_{1}, \ldots, i_{d}} \prod_{r=1}^{k} x_{i_{r}j_{r}} \prod_{l=1}^{m} g_{l_{r}}^{(l)} \right\|_{\|\cdot\|} \mid \forall r \leq k \sum_{l_{r}} \left( x_{i_{r}j_{r}}^{(r)} \right)^{2} \leq 1 \right\}, \quad (2.1)
\]

\[
\|A\|_{\mathcal{P}} := \sup \left\{ \mathbb{E} \left\| \sum_{i_{1}, \ldots, i_{d}} a_{i_{1}, \ldots, i_{d}} \prod_{r=1}^{k} x_{i_{r}j_{r}} \prod_{l \in [d] \cup \mathcal{P}} g_{l_{r}}^{(l)} \right\|_{\|\cdot\|} \mid \forall r \leq k \sum_{l_{r}} \left( x_{i_{r}j_{r}}^{(r)} \right)^{2} \leq 1 \right\}, \quad (2.2)
\]

where \(G^{(l)} = (g_{l_{r}}^{(l)})_{i_{r}j_{r}}, l = 1, \ldots, m\) are independent arrays of i.i.d. standard Gaussian variables.

We do not exclude the situation that \(\mathcal{P}'\) or \(\mathcal{P}\) is an empty partition. If \(\mathcal{P}' = \emptyset\), then \(\|A\|_{\mathcal{P}', \mathcal{P}} = \|A\|_{\mathcal{P}}\) is defined in non-probabilistic terms. Another case when \(\|A\|_{\mathcal{P}', \mathcal{P}} = \|A\|_{\mathcal{P}}\) is when \(\mathcal{P}'\) consists of singletons only.

In particular for \(d = 3\) we have (note that to shorten the notation we suppress some brackets and write e.g. \(\|A\|_{\{2\}, \{3\}}\) and \(\|A\|_{\{1\}} |_{\{2\}, \{3\}}\), instead of \(\|A\|_{\{1\}|_{\{2\}, \{3\}}}\) and
Theorem 2.1. Assume that $A = (a_{i_1, \ldots, i_d})_{i_1, \ldots, i_d}$ is a finite matrix with values in a Banach space $(F, \|\|)$. Then for any $p \geq 1$,

$$
\frac{1}{C(d)} \sum_{J \subset [d]} \sum_{\{P \in P(J)\}} p^{|P|/2} \|A\|_p \leq \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\|_p \\
\leq C(d) \sum_{(\mathcal{P}, \mathcal{P}') \in \mathcal{P}([d])} p^{|\mathcal{P}|/2} \|A\|_{P'} \|_{P}.
$$

(2.3)

The lower bound in (2.3) motivates the following conjecture (we leave it to the reader to verify that in general Banach spaces it is impossible to reverse the upper bound even for $d = 2$).

Conjecture 2.2. Under the assumption of Theorem 2.1 we have

$$
\left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\|_p \leq C(d) \sum_{J \subset [d]} \sum_{\{P \in P(J)\}} p^{|P|/2} \|A\|_p.
$$

(2.4)

Example 2.3. In particular for $d = 3$, Theorem 2.1 yields for symmetric matrices

$$
\frac{1}{C(S_1) \|A\|_p} \|A\|_p \leq C(S_1 + S_2).
$$
where (we recall) $C$ is a numerical constant and

$$S_1 := E \left\| \sum_{i,j,k} a_{ijk} g_i^{(1)} g_j^{(2)} g_k^{(3)} \right\| + p^{1/2} \sup_{\|x\| \leq 1} E \left\| \sum_{i,j,k} a_{ijk} g_i^{(1)} g_j^{(2)} x_k \right\|
$$

$$+ \sup_{\|x\| \leq 1} E \left\| \sum_{i,j,k} a_{ijk} g_i^{(1)} x_{jk} \right\|
$$

$$+ \sup_{\|x\| \leq 1} E \left\| \sum_{i,j,k} a_{ijk} x_{ijk} \right\|
$$

$$+ p \left( \sup_{\|x\| \leq 1, \|y\| \leq 1} E \left\| \sum_{i,j,k} a_{ijk} g_i^{(1)} x_{jiy} \right\| + \sup_{\|x\| \leq 1, \|y\| \leq 1} \left\| \sum_{i,j,k} a_{ijk} x_{iyy} \right\| \right)
$$

$$+ p^{3/2} \sup_{\|x\| \leq 1, \|y\| \leq 1, \|z\| \leq 1} \left\| \sum_{i,j,k} a_{ijk} x_{ijy} z_k \right\| ,$$

$$S_2 := E \left\| \sum_{i,j,k} a_{ijk} g_{ijk}^{(1)} \right\| + E \left\| \sum_{i,j,k} a_{ijk} g_{ijk}^{(1)} g_k^{(2)} \right\| + p^{1/2} \sup_{\|x\| \leq 1} E \left\| \sum_{i,j,k} a_{ijk} g_{ijk}^{(1)} \right\| .$$

**Remark 2.4.** Unfortunately we are able to show (2.4) only for $d = 2$ and with an additional factor $\ln p$ (cf. [2]). It is likely that by a modification of our proof one can show (2.4) for arbitrary $d$ with an additional factor $(\ln p)^{C(d)}$.

By a standard application of Chebyshev’s and Paley-Zygmund inequalities, Theorem 2.1 can be expressed in terms of tails.

**Theorem 2.5.** Under the assumptions of Theorem 2.1 the following two inequalities hold. For any $t > C(d) \sum_{p' \in P(d)} \|A\|_{\mathcal{P}'}$

$$\mathbb{P} \left( \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \ldots g_{i_d}^{(d)} \right\| \geq t \right) \leq 2 \exp \left( -\frac{1}{C(d)} \min_{|P'| > 0} \left( \frac{t}{\|A\|_{\mathcal{P}'}} \right)^{2/|P'|} \right) ,$$

and for any $t \geq 0$,

$$\mathbb{P} \left( \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \ldots g_{i_d}^{(d)} \right\| \geq \frac{1}{C(d)} \mathbb{E} \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \ldots g_{i_d}^{(d)} \right\| + t \right)
$$

$$\geq \frac{1}{C(d)} \exp \left( -C(d) \min_{\emptyset \neq J \subseteq [d]} \min_{P \in P(J)} \left( \frac{t}{\|A\|_{\mathcal{P}'}} \right)^{2/|P'|} \right) .$$

In view of (1.5) and [15] it is clear that to prove Theorem 2.1 one needs to estimate suprema of some Gaussian processes. The next statement is the key element of the proof of the upper bound in (2.3).

**Theorem 2.6.** Under the assumptions of Theorem 2.1 we have for any $p \geq 1$,

$$E \sup_{(x^{(2)}, \ldots, x^{(d)}) \in (B^d_p)^{d-1}} \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_i \prod_{k=2}^d x_{ik}^{(k)} \right\| \leq C(d) \sum_{(P, P') \in P(d)} p^{\frac{|P| + 1 + d}{2}} \|A\|_{\mathcal{P}'} .$$

(2.5)

We postpone proofs of the above results until Section 5 and discuss now some of their consequences.
2.1 Two-sided estimates in special classes of Banach spaces

We start by introducing a class of Banach spaces for which the estimate (2.3) is two-sided. To this end we restrict our attention to Banach spaces \((F, \|\cdot\|)\) which satisfy the following condition: there exists a constant \(K = K(F)\) such that for any \(n \in \mathbb{N}\) and any matrix \(\langle b_{ij} \rangle_{i,j \leq n}\) with values in \(F\),

\[
E \left\| \sum_{ij} b_{ij} g_{i,j} \right\| \leq K E \left\| \sum_{ij} b_{ij} g_{i,j}^\prime \right\|. \tag{2.6}
\]

This property appears in the literature under the name Gaussian property (\(\alpha+\)) (see [20]) and is closely related to Pisier’s contraction property [25]. It has found applications, e.g., in the theory of stochastic integration in Banach spaces. We refer to [11, Chapter 7] for a thorough discussion and examples, mentioning only that (2.6) holds for Banach spaces of type 2, and for Banach lattices (2.6) is equivalent to finite cotype.

**Remark 2.7.** By considering \(n = 1\) it is easy to see that \(K \geq \sqrt{\pi/2} > 1\).

A simple inductive argument and (2.6) yield that for any \(d, n \in \mathbb{N}\) and any \(F\)-valued matrix \(\langle b_{i_1,\ldots,i_d} \rangle_{i_1,\ldots,i_d \leq n}\),

\[
E \left\| \sum_{i} b_{i} g_{i} \right\| \leq K^{d-1} E \left\| \sum_{i} b_{i} g_{i}^{(1)} \cdots g_{i}^{(d)} \right\|, \tag{2.7}
\]

where we recall that \(i = (i_1, \ldots, i_d)\) and each \(i_1, \ldots, i_d\) runs through \([n]\). It turns out that under the condition (2.6) our bound (2.3) is actually two-sided.

**Proposition 2.8.** Assume that \((F, \|\cdot\|)\) satisfies (2.6) and \((P, P') \in \mathcal{P}([d])\). Then

\[
\| A \|_{P' \mid P} \leq K^{1 + |P'|-|P|} A \|_{P}.
\]

**Proof.** Let \(P' = (J_1, \ldots, J_k), P = (I_1, \ldots, I_m)\). Then \(|P'| - |P'| = \sum_{i=1}^{k} (|J_i| - 1)\). The proof is by induction on \(s := \{|l : |J_l| \geq 2|\}\). If \(s = 0\) the assertion follows by the definition of \(\|A\|_{P}\). Assume that the statement holds for \(s\) and \(|\{l : |J_l| \geq 2|\} = s + 1\). Without loss of generality \(|J_1| \geq 2\). Combining Fubini’s Theorem with (2.7) we obtain

\[
\| A \|_{P' \mid P} = \sup \left\{ E\left(G^{(1)}, \ldots, G^{(m)}\right) \left\| \sum_{i} a_i \prod_{r=1}^{m} x_{i}^{(r)} g_{i_1}^{(1)} \prod_{r=2}^{k} g_{i_2}^{(r)} \prod_{r=2}^{k} g_{i_2}^{(r)} \right\| \right\}
\leq K^{1 + |J_1| - 1} \sup \left\{ E\left(G^{(1)}, \ldots, G^{(m)}\right) \left\| \sum_{i} a_i \prod_{r=1}^{m} x_{i}^{(r)} \prod_{j \in J_1} \prod_{r=2}^{k} g_{i_1}^{(r)} \prod_{r=2}^{k} g_{i_2}^{(r)} \right\| \right\}
\leq K^{1 + |P'|-|P|} \| A \|_{P},
\]

where \(G^{(i)} = (g_{i_1}^{(j)})_{i_1, j \in J_1, i_2}\) are independent families of i.i.d. \(\mathcal{N}(0, 1)\) random variables and in the last inequality we used (conditionally) the induction assumption.  

The following corollary is an obvious consequence of Proposition 2.8 and Theorems 2.1, 2.5.

**Corollary 2.9.** For any Banach space \((F, \|\cdot\|)\) satisfying (2.6) we have for \(p \geq 1\),

\[
\left\| \sum_{i_1,\ldots,i_d} a_{i_1,\ldots,i_d} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\|_{p} \leq C(d) K^{d-1} \sum_{I \subset [d]} \sum_{P \in \mathcal{P}(I)} |P|^{p/2} \| A \|_{P},
\]
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and for \( t > C(d)K^{d-1}E\left\| \sum_{i_1} a_{i_1} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\| \),

\[
P \left( \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\| \geq t \right) \leq 2 \exp \left( -\frac{1}{C(d)}K^{2-2d}\eta(t) \right),
\]

where

\[
\eta(t) := \min_{\emptyset \neq I \subset [d]} \min_{P \in \mathcal{P}(I)} \left( \frac{t}{\|A\|_F} \right)^{2/|P|}.
\]

Thanks to infinite divisibility of Gaussian variables, the above corollary can be in fact generalized to arbitrary polynomials in Gaussian variables, as stated in the following theorem.

**Theorem 2.10.** Let \( F \) be a Banach space. If \( G \) is a standard Gaussian vector in \( \mathbb{R}^n \) and \( f: \mathbb{R}^n \to F \) is a polynomial of degree \( D \), then for all \( p \geq 2 \),

\[
\|f(G) - Ef(G)\|_p \geq \frac{1}{C(D)} \left( E\|f(G) - Ef(G)\| + \sum_{1 \leq d \leq D} \sum_{\emptyset \neq I \subset [d]} \sum_{P \in \mathcal{P}(I)} p^{d |P|} \|E\nabla^d f(G)\|_p \right) \tag{2.8}
\]

and for all \( t > 0 \),

\[
P \left( \|f(G) - Ef(G)\| \geq \frac{1}{C(D)} \left( E\|f(G) - Ef(G)\| + t \right) \right) \geq \frac{1}{C(D)} \exp \left( -C(D)\eta_f(t) \right), \tag{2.9}
\]

where

\[
\eta_f(t) = \min_{1 \leq d \leq D} \min_{\emptyset \neq I \subset [d]} \min_{P \in \mathcal{P}(I)} \left( \frac{t}{\|E\nabla^d f(G)\|_p} \right)^{2/|P|}.
\]

Moreover, if \( F \) satisfies (2.6), then for all \( p \geq 1 \),

\[
\|f(G) - Ef(G)\|_p \leq C(D)K^{D-1} \left( E\|f(G) - Ef(G)\| + \sum_{1 \leq d \leq D} \sum_{\emptyset \neq I \subset [d]} \sum_{P \in \mathcal{P}(I)} p^{d |P|} \|E\nabla^d f(G)\|_p \right) \tag{2.10}
\]

and for all \( t \geq C(D)K^{D-1}E\|f(G) - Ef(G)\|_p \),

\[
P \left( \|f(G) - Ef(G)\| \geq t \right) \leq 2 \exp \left( -C(D)^{-1}K^{2-2D}\eta_f(t) \right). \tag{2.11}
\]

The above theorem is an easy consequence of results for homogeneous decoupled chaoses and the following proposition, the proof of which (as well as the proof of the theorem) will be presented in Section 5.

**Proposition 2.11.** Let \( F \) be a Banach space, \( G \) a standard Gaussian vector in \( \mathbb{R}^n \) and \( f: \mathbb{R}^n \to F \) be a polynomial of degree \( D \). Then for \( p \geq 1 \),

\[
\|f(G) - Ef(G)\|_p \sim_D \sum_{d=1}^D \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d}^{(d)} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\|_p,
\]

where the \( d \)-indexed \( F \)-valued matrices \( A_d = (a_{i_1, \ldots, i_d}^{(d)})_{1 \leq i_1, \ldots, i_d \leq n} \) are defined as \( A_d = E\nabla^d f(G) \).
Remark 2.12. Let us stress that Proposition 2.11 as well as inequalities (2.8) and (2.9) of Theorem 2.10 hold in arbitrary Banach spaces. The assumption (2.6) is needed for inequalities (2.10) and (2.11). A positive answer to Conjecture 2.2 would allow to eliminate this assumption and remove the constant $K$ from the inequalities.

2.2 $L^q$ spaces

It turns out that $L^q$ spaces satisfy (2.6) and as a result upper and lower bounds in (2.3) are comparable. Moreover, as is shown in Lemma 2.14 below, in this case one may express all the parameters without any expectations. For the sake of brevity, we will focus on moment estimates, clearly tail bounds follow from them by standard arguments (cf. the proof of Theorem 2.5).

Proposition 2.13. For $q \geq 1$ the space $L^q(V, \mu)$ satisfies (2.6) with $K = C \sqrt{q}$.

Proof. From [11, Theorem 7.1.20] it follows that if $F$ is of type 2 with constant $T_2$, then it satisfies (2.6) with $K = T_2$, while it is well known that the type 2 constant of $L^q(V, \mu)$ is of order $\sqrt{q}$.

For a multi-indexed matrix $A$ of order $d$ with values in $L^q(V, \mu)$ and $J \subseteq [d]$, $\mathcal{P} = (I_1, \ldots, I_k) \in \mathcal{P}([J])$ we define

$$
\|A\|_{L^q, \mathcal{P}} = \sup \left\{ \left\| \sum_{i \in [d]\setminus J} \left( \sum_{j \in \mathcal{P}} a_{i_j} \prod_{r=1}^d \langle g^{(r)}_{i_r} \prod_{r=1}^d x^{(r)}_{i_{r_r}} \rangle \right)^{q} \right\|_{L^q} \right\}.
$$

For $J = [d]$ and $\mathcal{P} \in \mathcal{P}([d])$ we obviously have $\|A\|_{L^q, \mathcal{P}} = \|A\|_\mathcal{P}$. The following lemma asserts that for general $J$ the corresponding two norms are comparable.

Lemma 2.14. For any $J \subseteq [d]$, $\mathcal{P} = (I_1, \ldots, I_k) \in \mathcal{P}(J)$ and any multi-indexed matrix $A$ of order $d$ with values in $L^q(V, \mu)$ we have

$$
C(d)^{-1} q^{\frac{1-d+|J|}{2}} \|A\|_{L^q, \mathcal{P}} \leq \|A\|_\mathcal{P} \leq C(d) q^{\frac{|J|}{d}} \|A\|_{L^q, \mathcal{P}}.
$$

Proof. By Jensen’s inequality and Corollary A.7 we get

$$
\|A\|_\mathcal{P} \leq \sup \left\{ \left( \int_V \left| \sum_{i \in [d]\setminus J} \left( \prod_{j \in \mathcal{P}} g^{(j)}_{i_j} \prod_{r=1}^d x^{(r)}_{i_{r_r}} \right)^{q} \right| d\mu(v) \right)^{1/q} \right\} \leq C(d) q^{\frac{|J|}{d}} \sup \left\{ \left\| \sum_{i \in [d]\setminus J} \left( \prod_{j \in \mathcal{P}} a_{i_j} \prod_{r=1}^d x^{(r)}_{i_{r_r}} \right)^{q} \right\|_{L^q} \right\}.
$$

On the other hand Theorem A.1 (applied with $p = 1$) and Corollary A.7 yield

$$
\|A\|_\mathcal{P} \geq \frac{q^{\frac{|J|}{d}}} {C(d)} \sup \left\{ \left( \prod_{j \in \mathcal{P}} g^{(j)}_{i_j} \prod_{r=1}^d x^{(r)}_{i_{r_r}} \right)^{q} \right\} \leq \frac{q^{\frac{|J|}{d}}} {C(d)} \sup \left\{ \left\| \sum_{i \in [d]\setminus J} \left( \prod_{j \in \mathcal{P}} a_{i_j} \prod_{r=1}^d x^{(r)}_{i_{r_r}} \right)^{q} \right\|_{L^q} \right\}.
$$

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Theorem 2.15. Let $q \geq 1$ and let $A = (a_{i_1,\ldots,i_d})_{i_1,\ldots,i_d}$ be a multi-indexed matrix with values in $L_q(V, \mu)$. Then for any $p \geq 1$ we have

$$
\frac{1}{C(d)} \frac{q^{1-d}}{d} \sum_{J \subset [d]} \sum_{P \in P(J)} p \frac{|P|}{|J|} \|A\|_{L_q,P} \leq \sum_{i_1,\ldots,i_d} a_{i_1,\ldots,i_d} g^{(1)}_{i_1} \cdots g^{(d)}_{i_d} \|\|_{p} 
\leq C(d) q^{d-\frac{d}{2}} \sum_{J \subset [d]} \sum_{P \in P(J)} p \frac{|P|}{|J|} \|A\|_{L_q,P}.
$$

Proof. This is an obvious consequence of Theorem 2.1, Corollary 2.9, Proposition 2.13 and Lemma 2.14.

Using Proposition 2.11 we can extend the above result to general polynomials.

Theorem 2.16. Let $G$ be a standard Gaussian vector in $\mathbb{R}^n$ and let $f : \mathbb{R}^n \to L_q(V, \mu)$ ($q \geq 1$) be a polynomial of degree $D$. Then for $p \geq 1$, we have

$$
\frac{1}{C(D)} \sum_{d=1}^{D} q^{1-d} \sum_{J \subset [d]} \sum_{P \in P(J)} p \frac{|P|}{|J|} \|\nabla^d f(G)\|_{L_q,P} \leq \|f(G) - Ef(G)\|_{P} 
\leq C(D) \sum_{d=1}^{D} q^{d-\frac{d}{2}} \sum_{J \subset [d]} \sum_{P \in P(J)} p \frac{|P|}{|J|} \|\nabla^d f(G)\|_{L_q,P}.
$$

Example 2.17. Consider a general polynomial of degree 3, i.e.,

$$
f(G) = \sum_{i,j,k=1}^{n} a_{ijk} g_i g_j g_k + \sum_{i,j=1}^{n} b_{ij} g_i g_j + \sum_{i=1}^{n} c_i g_i + d,
$$

where the coefficients $a_{ijk}, b_{ij}, c_i, d$ take values in a Banach space and the matrices $(a_{ijk})_{i,j,k}, (b_{ij})_{i,j}$ are symmetric. Then one checks that

$$
\nabla^1 f(G) = \left(c_i + 3 \sum_{j=1}^{n} a_{ijk}\right)_{i=1}^{n},
\nabla^2 f(G) = 2(b_{ij})_{i,j=1}^{n},
\nabla^3 f(G) = 3(a_{ijk})_{i,j,k=1}^{n}.
$$

2.3 Exponential variables

Theorem 2.15 together with Lemma A.8 allows us to obtain inequalities for chaoses based on i.i.d standard symmetric exponential random variables (i.e., variables with density $2^{-1} \exp(-|t|)$) which are denoted by $(E_j^{(i)})_{i,j \in \mathbb{N}}$ below. Similarly as in the previous section we concentrate only on the moment estimates.

Proposition 2.18. Let $A = (a_{i_1,\ldots,i_d})_{i_1,\ldots,i_d}$ be a matrix with values in $L_q(V, \mu)$. Then for any $p \geq 1$, $q \geq 2$ we have

$$
\left\| \sum_{i} a_i \prod_{k=1}^{d} E_{i_k}^{(i)} \right\|_{p} \leq \sum_{I \subset [d]} \sum_{J \subset [d]} \sum_{P \in P(J) \setminus (\{I,J\})} p \frac{|I|+|P|}{|J|} \max_{i_I} \|a_{i_1,\ldots,i_d} E_{i_1,\ldots,i_d}\|_{L_q,P}.
$$

One can take $C^{-1}(d) q^{1/2-d}$ in the lower bound and $C(d) q^{2d-1/2}$ in the upper bound.
Example 2.19. If \( d = 2 \) then Proposition 2.18 reads for a symmetric matrix \( A = (a_{ij})_{i,j} \) as

\[
\left\| \sum_{ij} a_{ij} E^{(1)}_i E^{(2)}_j \right\|_p \sim p^2 \max_{i,j} \|a_{ij}\|_{L^q} + p^{3/2} \max_j \sum_i a_{ij} x_i \left\| L^q \right\|,
\]

\[
+ p \left( \max_{x \in B^2} \left\| \sum_{ij} a_{ij} x_i y_j \right\|_{L^q} + \max_j \left( \sum_i a^2_{ij} \right)^{1/2} \left\| L^q \right\| \right),
\]

\[
+ p^{1/2} \left( \sup_{x \in B^2} \left\| \sum_i \left( \sum_j a_{ij} x_j \right)^2 \right\|_{L^q} + \sup_{x \in B^2} \left\| \sum_{i,j} a_{ij} x_{ij} \right\|_{L^q} \right),
\]

\[
+ \left( \sum_{ij} a^2_{ij} \right)^{1/2} \left\| L^q \right\|. 
\]

The proof of Proposition 2.18 is postponed until Section 5.

3 Reformulation of Theorem 2.6 and entropy estimates

Let us rewrite Theorem 2.6 in a different language. As explained in the introduction, since the variables we consider take values in the finite dimensional subspace spanned by the coefficients \( a_i \), we may assume without loss of generality that \( F = \mathbb{R}^m \) for some finite \( m \) and \( a_{i_1, \ldots, i_d} = (a_{i_1, \ldots, i_d, i_{d+1}})_{i_{d+1} \leq m} \). For this reason from now on the multi-index \( i \) will take values in \( [n]^d \times [m] \) and all summations over \( i \) should be understood as summations over this set. Accordingly, the matrix \( A \) will be treated as a \( (d+1) \)-indexed matrix with real coefficients. Let \( T = B_{F^*} \) be the unit ball in the dual space \( F^* \) (where duality is realized on \( \mathbb{R}^m \) through the standard inner product). In the sequel we will therefore assume that \( T \) is a fixed nonempty symmetric bounded subset of \( \mathbb{R}^m \).

In this setup we have

\[
E \sup_{\langle x^{(2)}, \ldots, x^{(d)} \rangle \in (B^2)^{d-1}} \left\| \sum_i a_i g_i \prod_{k=2}^d x^{(k)}_{i_k} \right\| = E \sup_{\langle x^{(2)}, \ldots, x^{(d)} \rangle \in (B^2)^{d-1}} \sup_{t \in T} \sum_i a_i g_i \prod_{k=2}^d x^{(k)}_{i_k} t_{i_{d+1}},
\]

\[
\left\| A \right\|_{P' \mid P} = \sup \left\{ E \sup_{t \in T} \sum_i a_i \prod_{k=1}^r \prod_{l=1}^{s_k} g^{(l)}_{i_{k_l}} t_{i_{d+1}} \mid \forall k_1, \ldots, r \sum_{i_{k_1}} x^{(k)}_{i_{k_1}} = 1 \right\},
\]

(3.1)

where \( P = (I_1, \ldots, I_r), P' = (J_1, \ldots, J_s), (P', P) \in \mathcal{P}([d]) \).

To make the notation more compact we define

\[
s_k(A) = \sum_{(P, P') \in \mathcal{P}([d]) \atop |P|=k} \left\| A \right\|_{P' \mid P}.
\]

The next statement is a reformulation of Theorem 2.6 in the introduced setup. The proof of it will be presented in Section 4.

Theorem 3.1. For any \( p \geq 1 \) we have

\[
E \sup_{\langle x^{(2)}, \ldots, x^{(d)} \rangle \in (B^2)^{d-1} \times T} \sum_i a_i g_i \prod_{k=2}^d x^{(k)}_{i_k} t_{i_{d+1}} \leq C(d) \sum_{k=0}^{d-1} p^{k+1-d} s_k(A).
\]

(3.2)
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To estimate the supremum of a centered Gaussian process \((G_v)_{v \in V}\) one needs to study the distance on \(V\) given by
\[
d(v, v') := (E|G_v - G_{v'}|)^{1/2}
\]
(we refer to the monograph [28] for an extensive presentation of chaining techniques related to estimates of suprema of stochastic processes). In the case of the Gaussian process from (3.2) this distance is defined on \((B_2^n)^{d-1} \times T \subset \mathbb{R}^{n(d-1)} \times \mathbb{R}^m\) by the formula

\[
\rho_A((x_1^{(2)}, \ldots, x_1^{(d)}, t), (y_1^{(2)}, \ldots, y_1^{(d)}, t'))
\]
\[
= \left( \sum_{i_1} \sum_{i_2, \ldots, i_{d+1}} a_{i_1, \ldots, i_{d+1}} \left( \prod_{k=2}^d x_{i_k}^{(k)} t_{i_{k+1}} - \prod_{k=2}^d y_{i_k}^{(k)} t'_{i_{k+1}} \right)^2 \right)^{1/2}
\]

where \((\bigotimes_{k=2}^d x_{i_k}^{(k)}) \otimes t = (x_1^{(2)}, \ldots, x_1^{(d)}, t_i^{(d+1)})_{i_2, \ldots, i_{d+1}} \in \mathbb{R}^{n(d-1)}\) and \(a_A\) is a norm defined on \((\mathbb{R}^n) \otimes (\mathbb{R}^{d-1}) \otimes \mathbb{R}^m \simeq \mathbb{R}^{n(d-1)}\) given by

\[
a_A(x) := \left( \sum_{i_1} \left( \sum_{i_{d+1}} a_{i_1} x_{i_{d+1}} \right)^2 \right)^{1/2}.
\]

We will now provide estimates for the entropy numbers \(N(U, \rho_A, \varepsilon)\) for \(\varepsilon > 0\) and \(U \subset (B_2^n)^{d-1} \times T\) (recall that \(N(S, \rho, \varepsilon)\) is the minimal number of closed balls with diameter \(\varepsilon\) in metric \(\rho\) that cover the set \(S\)). To this end let us introduce some new notation. From now on \(G_n = (g_1, \ldots, g_n)\) and \(G_n^{(i)} = (g_1^{(i)}, \ldots, g_n^{(i)})\) stand for independent standard Gaussian vectors in \(\mathbb{R}^n\). For \(s > 0\), \(U = \{(x_1^{(2)}, \ldots, x_1^{(d)}, t) \in U\} \subset (\mathbb{R}^n)^{d-1} \times T\) we set

\[
W_d^U(\alpha_A, s) := \sum_{k=1}^{d-1} s^k \sum_{I \subset \{2, \ldots, d\} : |I| = k} W_I^U(\alpha_A),
\]

where

\[
W_I^U(\alpha_A) := \sup_{(x_1^{(2)}, \ldots, x_1^{(d)}, t) \in U} E\alpha_A \left( \left( \bigotimes_{k=2}^d (x_1^{(k)}(1 - 1_I^{(k)}) + G_1^{(k)}1_I^{(k)}) \right) \otimes t \right)
\]

We define a norm \(\beta_A\) on \((\mathbb{R}^n) \otimes (\mathbb{R}^{d-1}) \simeq \mathbb{R}^{n(d-1)}\) by (recall that we assume symmetry of the set \(T\))

\[
\beta_A(y) := \sup_{t \in T} \sum_{I} a_1g_1 y_{(d)\setminus(I)} t_{i_{d+1}} = \sup_{t \in T} \left| \sum_1 a_1g_1 y_{(d)\setminus(I)} t_{i_{d+1}} \right|
\]

Following (3.5) we denote

\[
V_d^U(\beta_A, s) := \sum_{k=0}^{d-1} s^{k+1} \sum_{I \subset \{2, \ldots, d\} : |I| = k} V_I^U(\beta_A),
\]

where

\[
V_I^U(\beta_A) := \sup_{(x_1^{(2)}, \ldots, x_1^{(d)}, t) \in U} E\beta_A \left( \left( \bigotimes_{k=2}^d (x_1^{(k)}(1 - 1_I^{(k)}) + G_1^{(k)}1_I^{(k)}) \right) \right)
\]
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Let us note that $V^U_d(\beta_A)$ depends on the set $U$ only through its projection on the first $d - 1$ coordinates.

We have

$$V^U_d(\beta_A, s) \geq s \cdot V^U_0(\beta_A) = s \cdot \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \beta_A \left( \bigotimes_{k=2}^d x^{(k)} \right). \quad (3.8)$$

Observe that by the classical Sudakov minoration (see Theorem A.2), for any $x^{(k)} \in \mathbb{R}^n$, $k = 2, \ldots, d$ there exists $T \subseteq x^{(k), \varepsilon} \subseteq T$ such that $|T \otimes x^{(k), \varepsilon}| \leq \exp(C\varepsilon^{-2})$ and

$$\forall t \in T \exists \varepsilon \in T \otimes x^{(k), \varepsilon} \alpha_A \left( \bigotimes_{k=2}^d x^{(k)} \otimes (t - t') \right) \leq \varepsilon \beta_A \left( \bigotimes_{k=2}^d x^{(k)} \right).$$

We define a measure $\mu^d_{\varepsilon,T}$ on $\mathbb{R}^{(d-1)n} \times T$ by the formula

$$\mu^d_{\varepsilon,T}(C) := \int_{\mathbb{R}^{(d-1)n}} \sum_{t \in T \otimes x^{(k), \varepsilon}} 1_C((x^{(2)}, \ldots, x^{(d)}, t)) d\gamma_{(d-1)n}(x^{(k)})_{k=2,\ldots,d},$$

where $\gamma_{n,t}$ is the distribution of $tG_n = t(g_1, \ldots, g_n)$. Clearly,

$$\mu^d_{\varepsilon,T}(\mathbb{R}^{(d-1)n} \times T) \leq \exp(C\varepsilon^{-2}). \quad (3.9)$$

To bound $N(U, \rho_A, \varepsilon)$ for $\varepsilon > 0$ and $U \subset (B_2^n)^{d-1} \times T$ we need two lemmas.

**Lemma 3.2 ([15, Lemma 2]).** For any $x = (x^{(1)}, \ldots, x^{(d)}) \in (B_2^n)^d$, norm $\alpha'$ on $\mathbb{R}^n$ and $\varepsilon > 0$ we have

$$\gamma_{dn, \varepsilon} \left( B_{\alpha'}(x, r(\varepsilon, \alpha')) \right) \geq 2^{-d} \exp(-d\varepsilon^{-2}/2),$$

where

$$B_{\alpha'}(x, r(\varepsilon, \alpha')) = \left\{ y = (y^{(1)}, \ldots, y^{(d)}) \in (\mathbb{R}^n)^d \mid \alpha' \left( \bigotimes_{k=1}^d x^{(k)} - \bigotimes_{k=1}^d y^{(k)} \right) \leq r(\varepsilon, \alpha') \right\}$$

and

$$r(\varepsilon, \alpha') = \sum_{k=1}^d \varepsilon^k \sum_{l \subset [d]} \sup_{|l| = k} \left( \bigotimes_{k=1}^d (x^{(k)}(1 - 1_{k \in l}) + C(l) 1_{k \in l}) \right).$$

**Lemma 3.3.** For any $(x, t) = (x^{(2)}, \ldots, x^{(d)}, t) \in (B_2^n)^{d-1} \times T$ and $\varepsilon > 0$ we have

$$\mu^d_{\varepsilon,T} \left( B \left( (x, t), \rho_A, W^d_{\varepsilon}(x, t) - (\beta_A, 8\varepsilon) \right) \right) \geq c^d \exp \left( -C(d)\varepsilon^{-2} \right).$$

**Proof.** Fix $(x, t) \in (B_2^n)^{d-1} \times T$, $\varepsilon > 0$ and consider

$$U = \left\{ (y^{(2)}, \ldots, y^{(d)}) \in \mathbb{R}^{(d-1)n} : \alpha_A \left( \bigotimes_{k=2}^d y^{(k)} - \bigotimes_{k=2}^d y^{(k)} \right) \right\}$$

$$+ \varepsilon \beta_A \left( \bigotimes_{k=2}^d x^{(k)} - \bigotimes_{k=2}^d y^{(k)} \right) \leq W^d_{\varepsilon}(x, t) - (\beta_A, 8\varepsilon) \right\}.$$

For any $(y^{(2)}, \ldots, y^{(d)}) \in U$ there exists $t' \in T \otimes y^{(k), \varepsilon}$ such that

$$\alpha_A \left( \bigotimes_{k=2}^d y^{(k)} \otimes (t - t') \right) \leq \varepsilon \beta_A \left( \bigotimes_{k=2}^d y^{(k)} \right).$$
By the triangle inequality,

\[
\alpha_A \left( \bigotimes_{k=2}^d x^{(k)} \otimes t - \bigotimes_{k=2}^d y^{(k)} \otimes t' \right) \\
\leq \alpha_A \left( \bigotimes_{k=2}^d x^{(k)} - \bigotimes_{k=2}^d y^{(k)} \otimes t \right) + \alpha_A \left( \bigotimes_{k=2}^d y^{(k)} \otimes (t - t') \right) \\
\leq \alpha_A \left( \bigotimes_{k=2}^d x^{(k)} - \bigotimes_{k=2}^d y^{(k)} \otimes t \right) + \varepsilon \beta_A \left( \bigotimes_{k=2}^d x^{(k)} - \bigotimes_{k=2}^d y^{(k)} \right) + \varepsilon \beta_A \left( \bigotimes_{k=2}^d x^{(k)} \right) \\
\leq W_d^{(x,t)}(\alpha_A, 4\varepsilon) + 2V_d^{(x,t)}(\beta_A, 4\varepsilon) \leq W_d^{(x,t)}(\alpha_A, 8\varepsilon) + V_d^{(x,t)}(\beta_A, 8\varepsilon),
\]

where in the third inequality we used (3.8). Thus,

\[
\mu^d_{\varepsilon,T} \left( B \left( (x,t), \rho_A, W_d^{(x,t)}(\alpha_A, 8\varepsilon) + V_d^{(x,t)}(\beta_A, 8\varepsilon) \right) \right) \geq \gamma(d-1)n,\varepsilon(U) \geq c^d \exp(-C(d)\varepsilon^{-2}),
\]

where the last inequality follows by Lemma 3.2 applied to the norm \( \alpha_A \cdot \otimes t + \varepsilon \beta_A(\cdot) \).

**Corollary 3.4.** Let \( U \subset (B_2^n)^{d-1} \times T \). Then for any \( \varepsilon > 0 \),

\[
N \left( U, \rho_A, W_d^U(\alpha_A, \varepsilon) + V_d^U(\beta_A, \varepsilon) \right) \leq \exp(C(d)\varepsilon^{-2})
\]

(3.10)

and for any \( \delta > 0 \),

\[
\sqrt{\log N(U, \rho_A, \delta)} \leq C(d) \left( \sum_{k=1}^{d-1} \left( \sum_{\iota \in \{2, \ldots, d\} \atop |\iota| = k} W_d^U(\alpha_A) \right)^{1/2} \delta^{-\frac{1}{2k}} + \sum_{k=0}^{d-2} \left( \sum_{\iota \in \{2, \ldots, d\} \atop |\iota| = k} V_d^U(\beta_A) \right)^{1/2} \delta^{-\frac{1}{2k+1}} \right).
\]

(3.11)

**Proof.** It suffices to show (3.10), since it easily implies (3.11). Consider first \( \varepsilon \leq 8 \). Obviously, \( W_d^U(\alpha_A, \varepsilon) + V_d^U(\beta_A, \varepsilon) \geq \sup_{(x,t) \in U} \left( W_d^{(x,t)}(\alpha_A) + V_d^{(x,t)}(\beta_A) \right) \). Therefore, by Lemma 3.3 (applied with \( \varepsilon/16 \)) we have for any \( (x,t) \in U \),

\[
\mu^d_{\varepsilon,T} \left( B \left( (x,t), \rho_A, W_d^U(\alpha_A, \varepsilon/2) + V_d^U(\beta_A, \varepsilon/2) \right) \right) \geq C(d)\varepsilon^{-1} \exp(-C(d)\varepsilon^{-2}).
\]

(3.12)

Suppose that there exist \( (x_1, t_1), \ldots, (x_N, t_N) \in U \) such that

\[
\rho_A((x_i, t_i), (x_j, t_j)) > W_d^U(\alpha_A, \varepsilon) + V_d^U(\beta_A, \varepsilon) \geq 2W_d^U(\alpha_A, \varepsilon/2) + 2V_d^U(\beta_A, \varepsilon/2)
\]

for \( i \neq j \).

Then the sets \( B \left( (x_i, t_i), \rho_A, W_d^U(\alpha_A, \varepsilon/2) + V_d^U(\beta_A, \varepsilon/2) \right) \) are disjoint, so by (3.9) and (3.12), we obtain \( N \leq C(d) \exp(C(d)\varepsilon^{-2}) \leq \exp(C(d)\varepsilon^{-2}) \).

If \( \varepsilon \geq 8 \) then (3.8) gives

\[
W_d^U(\alpha_A, \varepsilon) + V_d^U(\beta_A, \varepsilon) \geq 8 \sup_{(x(2), \ldots, x(d), t) \in U} E \left| \sum_{i} a_i g_i \prod_{k=2}^d x_i^{(k)} t_{i+1} \right| \geq \sqrt{\frac{128}{\pi}} \sup_{(x(2), \ldots, x(d), t) \in U} \left( \sum_{i_1} \sum_{i_2, \ldots, i_{d+1}} a_i \prod_{k=2}^d x_i^{(k)} t_{i+1} \right)^{1/2} \geq \text{diam} \left( U, \rho_A \right).
\]

So, \( N(U, \rho_A, W_d^U(\alpha_A, \varepsilon) + V_d^U(\beta_A, \varepsilon)) = 1 \leq \exp(\varepsilon^{-2}) \).
Remark 3.5. The classical Dudley’s bound on suprema of Gaussian processes (see e.g., [9, Corollary 5.1.6]) gives

\[ \mathbb{E} \left( \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in (B^d_2)^{d-1} \times T} \left| \sum_{i=1}^d a_i x_i(t) \right| \right) \leq C \int_0^\Delta \sqrt{\log N((B^d_2)^{d-1} \times T, \rho_A, \delta)} \, d\delta, \]

where \( \Delta \) is equal to the diameter of the set \((B^d_2)^{d-1} \times T\) in the metric \(\rho_A\). Unfortunately the entropy bound derived in Corollary 3.4 involves a nonintegrable term \(\delta^{-1}\). The remaining part of the proof of Theorem 3.1 is devoted to improving on Dudley’s bound.

For \(x, y \in (\mathbb{R}^n)^{d-1}\) we define a norm \(\hat{\alpha}_A\) on \((\mathbb{R}^n)^{d-1} = \mathbb{R}^{(d-1)n}\) by the formula

\[ \hat{\alpha}_A(x^{(2)}, \ldots, x^{(d)}) := \sup_{P \in \mathcal{P}([d]\setminus\{1\})} \left| \sum_{j=1}^d a_j x^{(j)}_t \right|, \]

where \(\mathcal{P}\) is the collection of all partitions of \([d]\setminus\{1\}\) into \(d-1\) nonempty subsets.

\[ \hat{\alpha}_A(x^{(2)}, \ldots, x^{(d)}) = \sup_{P \in \mathcal{P}([d]\setminus\{1\})} \left| \sum_{j=1}^d a_j x^{(j)}_t \right| = \sup_{P \in \mathcal{P}([d]\setminus\{1\})} \left| \sum_{j=1}^d a_j x^{(j)}_t \right|. \]

Proposition 3.6. For any \(d \geq 3\), \(\varepsilon > 0\) and \(U \subset (B^d_2)^{d-1} \times T\),

\[ N\left( \mathbb{E} \left( U, \rho_A, \sum_{k=0}^{d-2} \varepsilon^{d-k} s_k(A) + \varepsilon \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \hat{\alpha}_A(x^{(2)}, \ldots, x^{(d)}) \right) \right) \leq \exp(C(d)\varepsilon^{-2}). \]

Proof. We will estimate the quantities \(W^U_d(\alpha_A, \varepsilon)\) and \(V^U_d(\beta_A, \varepsilon)\) appearing in Corollary 3.4.

Since \(U \subset (B^d_2)^{d-1} \times T\), Jensen’s inequality yields for \(I \subset \{2, \ldots, d\}\),

\[ W^U_I(\alpha_A) = \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \mathbb{E}_A \left( \left( \bigotimes_{k=2}^d \left( x^{(k)}(1 - 1_I(k)) + G^{(k)} 1_I(k) \right) \right) \otimes t \right) \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \mathbb{E} \left( \left( \bigotimes_{k=2}^d \left( x^{(k)}(1 - 1_I(k)) + G^{(k)} 1_I(k) \right) \right) \otimes t \right) \]

\[ = \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \mathbb{E} \left( \left( \bigotimes_{k=2}^d \left( x^{(k)}(1 - 1_I(k)) + G^{(k)} 1_I(k) \right) \right) \otimes t \right) \]

\[ \leq \left| A \right|_{I \cup \{1\}, \{k\}; k \in \{d\} \setminus (I \cup \{1\})} \leq s_{d-1}(A). \quad (3.13) \]

By estimating a little more accurately in the second inequality in (3.13) we obtain for \(2 \leq j \leq d\),

\[ W^U_{\{j\}}(\alpha_A) \]

\[ \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \sum_{I : 2 \leq I \leq d} \sup_{y^{(2)}, \ldots, y^{(d)} \in (B^d_2)^{d-1} \setminus I} \left( \sum_{i_1, j_t} \left( \sum_{1 \leq k \leq d} a_{i_1} x^{(i_1)}_t \prod_{k \in I \setminus \{1, j\}} y^{(k)}_t \right)^2 \right) \]

\[ \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \sum_{I : 2 \leq I \leq d} \left( \sum_{i_1, j_t} \left( \sum_{1 \leq k \leq d} a_{i_1} x^{(i_1)}_t \prod_{k \in I \setminus \{1, j\}} y^{(k)}_t \right)^2 \right) \]

\[ \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \sum_{I : 2 \leq I \leq d} \left( \sum_{i_1, j_t} \left( \sum_{1 \leq k \leq d} a_{i_1} x^{(i_1)}_t \prod_{k \in I \setminus \{1, j\}} y^{(k)}_t \right)^2 \right). \quad (3.14) \]
Observe that (3.14) is not true for $d = 2$ (cf. Remark 3.7).

Let us now pass to the quantity $V^U_d (\beta, \varepsilon)$. The definition of $V^U_d$ and the inclusion $U \subset (B^2_d)^{d-1} \times T$ yield
\[ V^U_d (\beta) \leq \| A \|_1 (1: i \in I \cap \{ k : k \in [d] \cap I \cup 1 \}) \leq s_{d-|I|-1} (A) \text{ for } I \neq \emptyset \quad (3.15) \]
and
\[ V^U_d (\beta) \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \sum_{l=2}^{d} \sup_{(y^{(2)}, \ldots, y^{(d)}) \in (B^2_d)^{d-1}} E \sup_{t' \in T} \sum_{i} a_{l} g_{i} x^{(l)}_{i} \prod_{2 \leq k \leq d, k \neq l} y^{(k)}_{i} t_{k+1} \]
\[ \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \sum_{l=2}^{d} \left \| \sum_{i} a_{l} x^{(l)}_{i} \right \|_{1} \left \| \{ 1 \} \cap \{ k : k \in [d] \cap 1 \} \right \}. \quad (3.16) \]

Inequalities (3.13)–(3.16) imply that
\[ W^d_d (\alpha, \varepsilon) + V^U_d (\beta, \varepsilon) \]
\[ = \sum_{k=2}^{d} \varepsilon^k \sum_{l \subset \{ 2, \ldots, d \} : |l| = k} W^U_d (\alpha) + \sum_{k=1}^{d-1} \varepsilon^{k+1} \sum_{l \subset \{ 2, \ldots, d \} : |l| = k} V^U_d (\beta) \]
\[ + \varepsilon \left ( \sum_{j=2}^{d} W^U_{j} (\alpha) + V^U_{j} (\beta) \right ) \]
\[ \leq C(d) \sum_{k=0}^{d-2} \varepsilon^d s_k (A) + d \varepsilon \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \left ( \sum_{l=2}^{d} \frac{\sum_{l' \in P} \sum_{l' \in P} \left \| \sum_{i} a_{l} x^{(l)}_{i} \right \|_{p' \in p} \right ) \]}

Hence the assertion is a simple consequence of Corollary 3.4. \qed

**Remark 3.7.** Proposition 3.6 is not true for $d = 2$. The problem arises in (3.14) – for $d = 2$ there does not exist $P \in \mathcal{P}([d] \cap 1)$ such that $|P| = d - 2$. This is the main reason why proofs for chaoses of order $d = 2$ (cf. [2]) have a different nature than for higher order chaoses.

### 4 Proof of Theorem 3.1

We will prove Theorem 3.1 by induction on $d$ (recall that the matrix $A$ has order $d+1$). To this end we need to amplify the induction hypothesis. For $U \subset (R^n)^{d-1} \times R^m$ we define
\[ F_A (U) = E \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \sum_{l=1}^{d} a_{l1, \ldots, l_{d+1}} g_{l1} \prod_{k=2}^{d} x^{(k)}_{l_{k+1}}. \]

**Theorem 4.1.** For any $U \subset (B^2_d)^{d-1} \times T$ and any $p \geq 1$
\[ F_A (U) \leq C(d) \left ( \sqrt{p} \Delta_A (U) + \sum_{k=0}^{d-1} p^{k+1-d} s_k (A) \right ), \quad (4.1) \]
where
\[ \Delta_A (U) = \sup_{(x^{(2)}, \ldots, x^{(d)}, t), (y^{(2)}, \ldots, y^{(d)}, t') \in U} \rho_A ((x^{(2)}, \ldots, x^{(d)}, t), (y^{(2)}, \ldots, y^{(d)}, t')) \]
\[ = \text{diam} (A, \rho_A). \]
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Clearly it is enough to prove Theorem 4.1 for finite sets $U$. Observe that

$$\Delta_n((B^2_{2d})^{d-1} \times T) \leq 2 \|A\|_{(j): j \in [d]} = 2s_d(A),$$

thus Theorem 4.1 implies Theorem 3.1. We will prove (4.1) by induction on $d$, but first we will show several consequences of the theorem. In the next three lemmas, we shall assume that Theorem 4.1 (and thus also Theorem 3.1) holds for all matrices of order smaller than $d + 1$.

**Lemma 4.2.** Let $p \geq 1$, $l \geq 0$ and $d \geq 3$. Then

$$N \left( \left( B_{2d} \right)^{d-1}, \hat{A}, 2^{-l} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A) \right) \leq \exp(C(d)2^{2l}p),$$

where $\hat{A}$ is the distance on $(\mathbb{R}^n)^{d-1}$ corresponding to the norm $\hat{A}$.

**Proof.** Note that

$$E\hat{A} \left( G^{(2)}, \ldots, G^{(d)} \right) = \sum_{j=2}^{d} \sum_{(P, P') \in \mathcal{P}[d] \setminus \{(j)\}} E \left\| \sum_{i \in j} a_i g_{ij} \right\|_{P \mid P}.$$

Up to a permutation of the indexes we have two possibilities

$$\left\| \sum_{i \in j} a_i g_{ij} \right\|_{P \mid P} = \left\| \sum_{i \in j} a_i g_{ij} \right\|_{\emptyset \{1, 2\}, \{l\} : 3 \leq l \leq d, l \neq j} \quad \text{or} \quad \left\| \sum_{i \in j} a_i g_{ij} \right\|_{\{1\} \{l\} : 2 \leq l \leq d, l \neq j}.$$  

First assume that

$$\left\| \sum_{i \in j} a_i g_{ij} \right\|_{\emptyset \{1, 2\}, \{l\} : 3 \leq l \leq d, l \neq j} = \left\| \sum_{i \in j} b_{i_1 \ldots i_d} g_{i_1} \right\|_{\emptyset \{2\} \ldots \{d-1\}}.$$

for an appropriately chosen matrix $B = (b_{i_1 \ldots i_d})$ (we treat a pair of indices $\{1, 2\}$ as a single index and renumber the indices in such a way that $j$, $\{1, 2\}$ and $d + 1$ would become $1$, $2$ and $d$ respectively).

Clearly,

$$\sum_{(P, P') \in \mathcal{P}[d-1]} \left\| B \right\|_{P \mid P} = \sum_{(P, P') \in \mathcal{C}[d]} \left\| A \right\|_{P \mid P} \leq \sum_{(P, P') \in \mathcal{C}[d]} \left\| A \right\|_{P \mid P} = s_k(A),$$

where $\mathcal{C} \subset \mathcal{P}[d]$ is the set of partitions which do not separate 1 and 2.

Thus, Theorem 3.1 applied to the matrix $B$ of order $d$ yields

$$E \left\| \sum_{i \in j} a_i g_{ij} \right\|_{\emptyset \{1, 2\}, \{l\} : 3 \leq l \leq d, l \neq j} = E \left\| \sum_{i_1} b_{i_1 \ldots i_d} g_{i_1} \right\|_{\emptyset \{2\} \ldots \{d-1\}} \leq C(d) \sum_{(P, P') \in \mathcal{P}[d-1]} \left\| B \right\|_{P \mid P} \leq C(d) \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A).$$
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Now assume that \( \| \sum_{ij} a_{ij} g_{ij} \|_{p^*} = \| \sum_{ij} a_{ij} g_{ij} \|_{\{1\} \times \{2 \leq i \leq d, l \neq j\}} \) and observe that

\[
E \left\| \sum_{ij} a_{ij} g_{ij} \right\|_{\{1\} \times \{2 \leq i \leq d, l \neq j\}} = E^g \left( \sup_{x^{(t)} \in B_2^n, 2 \leq t \leq d, l \neq j} \sup_{i \in T} \sum_{t=1}^{k} a_{ij} g_{ij} \prod_{2 \leq t \leq d, l \neq j} x^{(t)}_{i,l+1} \right) = E \left( \sup_{x^{(t)} \in B_2^n, 2 \leq t \leq d, l \neq j} \prod_{2 \leq t \leq d, l \neq j} x^{(t)}_{i,l+1} \right) = E \left( \sup_{x^{(t)} \in B_2^n, 2 \leq t \leq d, l \neq j} \sum_{m \in M} d_{i_1, \ldots, i_d} g_{ij} \prod_{l=2}^{d-1} x^{(l)}_{i_l} \right),
\]

where \( D = (d_{i_1, \ldots, i_d})_{i_1, \ldots, i_d} \) is an appropriately chosen matrix of order \( d \), the symmetric set \( M \subset \mathbb{R}^n \) satisfies

\[E \sup_{t \in T} \sum_{i,j} b_{i,j} t_j = \sup_{m \in M} \sum_{i,j} b_{i,j} m_{i,j} \text{ for any matrix } (b_{i,j})_{i \leq j \leq m},\]

and \( \tilde{M} \) corresponds to \( M \) under a natural identification of \( \mathbb{R}^n \otimes \mathbb{R}^m \) with \( \mathbb{R}^{nm} \).

Applying Theorem 3.1 to the matrix \( D \) of order \( d \) gives

\[
E \left\| \sum_{ij} a_{ij} g_{ij} \right\|_{\{1\} \times \{2 \leq i \leq d, l \neq j\}} \leq C(d) \sum_{P \in P(d-1)} p^{|P|+2-d} \| A \|_{p^*}^{\| \tilde{M} \|_{p^*}} \leq C(d) \sum_{P \in P(d)} p^{|P|+2-d} \| A \|_{p^*}^{\| \tilde{M} \|_{p^*}} \leq C(d) \sum_{k=0}^{d-1} p^{k+2-d} s_k(A),
\]

(4.6)

where \( \| A \|_{p^*}^{\| \tilde{M} \|_{p^*}} \) is defined in the same manner as \( \| A \|_{p^*} \) (see (3.1)) but the supremum is taken over the set \( \tilde{M} \) instead of \( T \). The second inequality in (4.6) can be justified analogously as (4.4).

Combining (4.2), (4.3), (4.5), (4.6) and the dual Sudakov inequality (Theorem A.3, note that \( (B_2^n)^{d-1} \subset \sqrt{d-1} B_2^n \)) we obtain

\[
N \left( (B_2^n)^{d-1}, \hat{\rho}_A, t \sum_{k=0}^{d-1} p^{k+2-d} s_k(A) \right) \leq N \left( (B_2^n)^{d-1}, \hat{\rho}_A, C(d)^{-1} t E \hat{\rho}_A (G^{(2)}, \ldots, G^{(d)}) \right) \leq \exp(C(d)t^{-2}).
\]

It is now enough to choose \( t = (\sqrt{d})^{-1} \).

From now on for \( U \subset (\mathbb{R}^n)^{d-1} \times \mathbb{R}^n \) we denote

\[
\hat{\rho}_A(U) = \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \hat{\rho}_A(x^{(2)}, \ldots, x^{(d)}).
\]

Lemma 4.3. Suppose that \( d \geq 3 \), \( y = (y^{(2)}, \ldots, y^{(d)}) \in (B_2^n)^{d-1} \) and \( U \subset (B_2^n)^{d-1} \times T \). Then for any \( p \geq 1 \) and \( l \geq 0 \), we can find a decomposition

\[
U = \bigcup_{j=1}^{N} U_j, \quad N \leq \exp(C(d)2^dp)
\]
such that for each \( j \leq N \),

\[
F_A((y,0) + U_j) \leq F_A(U_j) + C(d) \left( \hat{\alpha}_A(y) + \hat{\alpha}_A(U) + 2^{-d} \sum_{k=0}^{d-2} \rho \frac{k+1-d}{2} s_k(A) \right) \tag{4.7}
\]

and

\[
\Delta_A(U_j) \leq 2^{-l} p^{-1/2} \hat{\alpha}_A(U) + 2^{-2l} \sum_{k=0}^{d-2} \rho \frac{k+d}{2} s_k(A). \tag{4.8}
\]

**Proof.** Fix \( y \in (B^2_2)^{d-1} \) and \( U \subset (B^2_2)^{d-1} \times T \). For \( I \subset \{2, \ldots, d\} \), \( x = (x^{(2)}, \ldots, x^{(d)}, t) \), \( \tilde{x} = (\tilde{x}^{(2)}, \ldots, \tilde{x}^{(d)}, t') \in (R^n)^{d-1} \times R^m \) and \( S \subset (R^n)^{d-1} \times R^m \), we define

\[
\rho_{A,I}^x(y, \tilde{x}) := \left[ \sum_{i_1} \left( \sum_{i_2, \ldots, i_{d+1}} a_{i_1} \prod_{k \in I} y^{(k)}_{i_k} \prod_{j \leq i_{d+1} \in I} x^{(j)}_{t_{i_{d+1}}} \prod_{j \in I} \tilde{x}^{(j)}_{t_{i_{d+1}}} \right) \right]^2,
\]

\[
\Delta_{A,I}^x(S) := \sup \left\{ \rho_{A,I}^x(y, \tilde{x}) : x, \tilde{x} \in S \right\}
\]

and

\[
F_{A,I}^x(S) := E \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in S} \sum_{1} a_1 g_1 \prod_{k \in I} y^{(k)}_{i_k} \prod_{j \leq i_{d+1} \in I} x^{(j)}_{t_{i_{d+1}}}.
\]

If \( I = \{2, \ldots, d\} \) then for \( S \subset (B^2_2)^{d-1} \times T \) we have

\[
F_{A,\{2, \ldots, d\}}^x(S) \leq E \sup_{t \in T} \sum_{1} a_1 g_1 d \prod_{k=2} y^{(k)}_{i_k} t_{i_{d+1}} \leq \hat{\alpha}_A(y) + \Delta_{A,I}^x(S).
\]

If \( 0 \neq I \subset \{2, \ldots, d\} \) then Theorem 4.1 applied to the matrix

\[
A(y, I) := \left( \sum_{i_{d+1}} a_1 \prod_{k \in I} y^{(k)}_{i_k} \right)_{I_{d+1}^{d-1}}
\]

of order \( d - |I| + 1 < d + 1 \) gives for any \( S \subset (B^2_2)^{d-1} \times T \) and \( q \geq 1 \),

\[
F_{A,I}^y(S) \leq C(d - |I|) \left( q^{1/2} \Delta_{A,I}^y(S) + \sum_{k=0}^{d-|I|-1} q^{k+1-d/2} s_k(A(y, I)) \right).
\]

For any \( 2 \leq k \leq d \), \( y^{(k)} \in B^2_2 \), thus \( s_k(A(y, I)) \leq s_{k+|I|}(A) \) for \( k < d - |I| - 1 \) and \( s_{d-|I|-1}(A(y, I)) \leq \hat{\alpha}_A(y) \).

Hence,

\[
F_{A,I}^y(S) \leq C(d - |I|) \left( q^{1/2} \Delta_{A,I}^y(S) + \hat{\alpha}_A(y) + \sum_{k=0}^{d-2} q^{k+1-d/2} s_k(A) \right). \tag{4.10}
\]
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By the triangle inequality,

$$F_A((y^{(2)}, \ldots, y^{(d)}, 0) + S) - F_A(S) \leq \sum_{\emptyset \neq I \subseteq \{2, \ldots, d\}} F^y_I(A).$$

Combining (4.9) and (4.10) we obtain for $S \subseteq (B_2^n)^{d-1} \times T$ and $q \geq 1$,

$$F_A((y^{(2)}, \ldots, y^{(d)}, 0) + S) \leq F_A(S) + C(d) \left( \hat{\alpha}_A(y) + \sum_{\emptyset \neq I \subseteq \{2, \ldots, d\}} q^{1/2} \Delta^y_I(S) + \sum_{k=0}^{d-2} q^k s_k(A) \right). \quad (4.11)$$

Fix $I \subsetneq \{2, \ldots, d\}$, $|I| < d - 2$ (we do not exclude $I = \emptyset$). Taking supremum over $y \in (B_2^n)^{d-1}$ we conclude that

$$\sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \hat{\alpha}_A(y_I)((x^{(k)}))_{k \in \{2, \ldots, d\} \setminus I} \leq \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \hat{\alpha}_A(x^{(2)}, \ldots, x^{(d)}).$$

Recall also that $s_k(A(y, I)) \leq s_{k+|I|}(A)$, thus we may apply $2^{d-1} - d$ times Proposition 3.6 with $\varepsilon = 2^{-1}p^{-1/2}$ and find a decomposition $U = \bigcup_{j=1}^{N_1} U_j$, $N_1 \leq \exp(C(d)2^k p)$ such that for each $j$ and $I \subset \{2, \ldots, d\}$ with $|I| < d - 2$,

$$\Delta^y_I(U_j) \leq 2^{-1}p^{-1/2} \hat{\alpha}_A(U) + 2^{-2l} \sum_{k=0}^{d-2} p^{k+\frac{d-2}{2}} s_k(A). \quad (4.12)$$

If $|I| = d - 2$ then the distance $\rho^y_{A,I}$ corresponds to a norm $\alpha_A(y,I)$ on $\mathbb{R}^m$ given by

$$\alpha_A(y,I)(x) = \left( \sum_{i_1} \left( \sum_{i_2, \ldots, i_{d+1}} a_{i_1} x_{i_1} y^{(k)}_{i_1} \prod_{k \in I} y^{(k)}_{i_k} \right)^2 \right)^{1/2},$$

where $j$ is defined by the condition $\{1, j\} = [d] \setminus I$ (cf. (3.3), (3.4) and observe that $A(y, I)$ is an $n \times m$ matrix). We define also (as in (3.6))

$$\beta_A(y,I)(x) = \mathbb{E} \sup_{t \in T} \sum_i a_i g_i x_i \prod_{l \in I} y^{(1)}_{i_l} t_{i_{d+1}}.$$

Recall the definitions (3.5) and (3.7) and note that (denoting by $\tilde{U}$ the projection of $U$ onto the $j$-th and $(d+1)$-th coordinate)

$$W^y_{\tilde{U}}(\alpha_A(y,I), \varepsilon) = \varepsilon \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \mathbb{E} \left[ \sum_{i_1} \left( \sum_{i_2, \ldots, i_{d+1}} a_{i_1} g_{i_1} t_{i_{d+1}} \prod_{k \in I} y^{(k)}_{i_k} \right)^2 \right]^{1/2},$$

$$\leq \varepsilon \sup_{(x^{(2)}, \ldots, x^{(d)}, t) \in U} \left[ \sum_{i_1, i_j} \left( \sum_{l_1, \ldots, l_{d+1}} a_{i_1} g_{i_1} t_{i_{d+1}} \right) \prod_{k \in I} y^{(k)}_{i_k} \right] \leq \varepsilon \hat{\alpha}(y), \quad (4.13)$$

where we again used that $y_k \in B_2^n$, $U \subset (B_2^n)^{d-1} \times T$.

We also have

$$V^y_{\{1\}}(\beta_A(y,I)) = \mathbb{E} \sup_{t \in T} \sum_i a_i g^{(1)}_{i_1} g^{(2)}_{i_2} \prod_{k \in I} y^{(k)}_{i_k} t_{i_{d+1}} \leq s_{d-2}(A).$$
and

\[ V_{1}^{\mathcal{I}}(\beta_{A}^{(y,t)}) = \sup_{(x^{(2)},...,(x^{(d)},t)) \in U} \mathbb{E} \sup_{i' \in I} \sum_{i} a_{i} g_{i} x_{i}^{(j)} \prod_{k \in I} y_{i}^{(k)} t_{i+k+1} \leq \hat{\alpha}(y). \]

Thus

\[ V_{1}^{\mathcal{I}}(\beta_{A}^{(y,t)}, \varepsilon) \leq \varepsilon \hat{\alpha}(y) + \varepsilon^{2} s_{d-2}(A). \]

Taking \( \varepsilon = 2^{-l-1} p^{-1/2} \) and combining the above estimate with (4.13) and Corollary 3.4 (applied \( d-1 \) times) we obtain a partition \( U = \bigcup_{j=1}^{N_{2}} U_{j}^{n} \) with \( N_{2} \leq \exp(C(d)2^{d}p) \) and

\[ \Delta_{A}^{\mathcal{I}}(U_{j}^{n}) \leq 2^{-l} p^{-1/2} \hat{\alpha}_{A}(y) + 2^{-2l} p^{-1} s_{d-2}(A) \quad (4.14) \]

for any \( I \subset \{2,...,d\} \) with \( |I| = d-2 \) and \( j \leq N_{2} \).

Intersecting the partition \( (U_{j}^{n})_{j \leq N_{1}} \) (which fulfills (4.12)) with \( (U_{j}^{n})_{j \leq N_{2}} \) we obtain a partition \( U = \bigcup_{i=1}^{N_{1}} U_{i} \) with \( N \leq N_{1} N_{2} \leq \exp(C(d)2^{d}p) \) and such that for every \( i \leq N \) there exist \( j \leq N_{1} \) and \( l \leq N_{2} \) such that \( U_{i} \subset U_{j}^{n} \cap U_{l}^{n} \).

Inequality (4.7) follows by (4.11) with \( q = 2^{2l} p \), (4.12) and (4.14). Observe that (4.8) follows by (4.12) for \( I = \emptyset \).

**Lemma 4.4.** Suppose that \( U \) is a finite subset of \( (B_{2}^{n})^{d-1} \times T \), with \( |U| \geq 2 \) and \( U-U \subset (B_{2}^{n})^{d-1} \times (T-T) \). Then for any \( p \geq 1, l \geq 0 \) there exist finite sets \( U_{i} \subset (B_{2}^{n})^{d-1} \times T \) and \( (y_{i}, t_{i}) \in U_{i}, i = 1, \ldots, N \) such that

(i) \( 2 \leq N \leq \exp(C(d)2^{d}p) \),

(ii) \( U = \bigcup_{i=1}^{N_{1}} ((y_{i}, 0) + U_{i} U_{i} \subset U-U, |U_{i}| \leq |U|-1, \)

(iii) \( \Delta_{A}(U_{i}) \leq 2^{-2l} \sum_{k=0}^{d-1} p^{k/d} s_{k}(A), \)

(iv) \( \hat{\alpha}_{A}(U_{i}) \leq 2^{-l} \sum_{k=0}^{d-1} p^{k+1/d} s_{k}(A), \)

(v) \( F_{A}((y_{i}, 0) + U_{i}) \leq F_{A}(U_{i}) + C(d) \left( \hat{\alpha}_{A}(U) + 2^{-l} \sum_{k=0}^{d-1} p^{k+1/d} s_{k}(A) \right). \)

**Proof.** By Lemma 4.2 we get

\[ (B_{2}^{n})^{d-1} = \bigcup_{i=1}^{N_{1}} B_{i}, N_{1} \leq \exp(C(d)2^{d}p), \]

where the diameter of the sets \( B_{i} \) in the norm \( \hat{\alpha} \) satisfies

\[ \text{diam}(B_{i}, \hat{\alpha}_{A}) \leq 2^{-l} \sum_{k=0}^{d-1} p^{k+1/d} s_{k}(A). \]

Let \( U_{i} = U \cap (B_{i} \times T) \). Selecting arbitrary \( (y_{i}, t_{i}) \in U_{i} \) (we can assume that these sets are nonempty) and using Lemma 4.3 (with \( l+1 \) instead of \( l \)) we decompose \( U_{i} - (y_{i}, 0) \) into \( \bigcup_{j=1}^{N_{2}} U_{ij} \) in such a way that \( N_{2} \leq \exp(C(d)2^{d}p) \),

\[ F_{A}((y_{i}, 0) + U_{ij} \leq F_{A}(U_{ij}) + C(d) \left( \hat{\alpha}_{A}(y_{i}) + \hat{\alpha}_{A}(U_{i} - (y_{i}, 0)) + 2^{-l} \sum_{k=0}^{d-2} p^{k+1/d} s_{k}(A) \right) \]

\[ \leq F_{A}(U_{ij}) + C(d) \left( \hat{\alpha}_{A}(y_{i}) + \text{diam}(B_{i}, \hat{\alpha}_{A}) + 2^{-l} \sum_{k=0}^{d-2} p^{k+1/d} s_{k}(A) \right) \]

\[ \leq F_{A}(U_{ij}) + C(d) \left( \hat{\alpha}_{A}(U) + 2^{-l} \sum_{k=0}^{d-1} p^{k+1/d} s_{k}(A) \right). \]
and
\[
\Delta_A(U_{ij}) \leq 2^{-l-1}p^{-1/2} \hat{\alpha}_A(U_i - (y_i, 0)) + 2^{-2l-2} \sum_{k=0}^{d-2} p^{\frac{k+d}{2}} s_k(A) \leq 2^{-2l} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A).
\]

We take the decomposition \( U = \bigcup_{i,j}((y_i,0) + U_{ij}) \). We have \( N = N_1N_2 \leq \exp(C(d)2^l p) \). Without loss of generality we can assume \( N \geq 2 \) and \( |U_{ij}| \leq |U|-1 \). Obviously, \( U_{ij} - U_{ij} \subset U_i - U \subset U - U \) and \( \hat{\alpha}_A(U_{ij}) \leq \hat{\alpha}_A(U_i - (y_i,0)) \leq 2^{-l} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A) \). A relabeling of the obtained decomposition concludes the proof. \( \square \)

**Proof of Theorem 4.1.** In the case of \( d = 2 \) Theorem 4.1 is proved in [2] (see Remark 37 therein).

Assuming (4.1) to hold for \( \{2, \ldots, d-1\} \), we will prove it for \( d \geq 3 \). Let \( U \subset (R^n)^{d-1} \times T \) and let us put \( \Delta_0 = \Delta_A(U) \), \( \hat{\Delta}_0 = \hat{\alpha}_A((B_{d-1}^n) \times T) \leq C(d)s_{d-1}(A) \),
\[
\Delta_l := 2^{2l-2} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A), \quad \hat{\Delta}_l := 2^{2l-1} \sum_{k=0}^{d-1} p^{\frac{k+1+d}{2}} s_k(A) \text{ for } l \geq 1.
\]

Suppose first that \( U \subset (\frac{1}{4}(B_{d-1}^n)) \times T \) and define
\[
c_U(r,l) := \sup \left\{ F_A(S) : S \subset (B_{d-1}^n) \times T, S \subset U - U, |S| \leq r, \Delta_A(S) \leq \Delta_l, \hat{\alpha}_A(S) \leq \hat{\Delta}_l \right\}.
\]
Note that any subset \( S \subset U \) satisfies \( \Delta_A(S) \leq \Delta_0 \) and \( \hat{\alpha}_A(S) \leq \hat{\Delta}_0 \), therefore,
\[
c_U(r,0) \geq \sup \{ F_A(S) : S \subset U, |S| \leq r \}. \quad (4.15)
\]
We will now show that for \( r \geq 2 \),
\[
c_U(r,l) \leq c_U(r - 1, l + 1) + C(d) \left( \hat{\Delta}_l + 2^l \sqrt{p} \Delta_l + 2^{-l} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A) \right). \quad (4.16)
\]
Indeed, let us take \( S \subset (B_{d-1}^n) \times T \) as in the definition of \( c_U(r,l) \). Then by Lemma 4.4 we may find a decomposition \( S = \bigcup_{i=1}^{N}((y_i,0) + S_i) \) satisfying (i)-(v) with \( U, U_i \) replaced by \( S, S_i \). Hence, by Lemma A.4, we have
\[
F_A(S) \leq C \sqrt{\log N} \Delta_A(S) + \max_i F_A((y_i,0) + S_i)
\]
\[
\leq C(d) \left( 2^{l} \sqrt{p} \Delta_l + \hat{\alpha}_A(S) + 2^{-l} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A) \right) + \max_i F_A(S_i). \quad (4.17)
\]
We have \( \Delta_A(S_i) \leq \Delta_{l+1}, \hat{\alpha}_A(S_i) \leq \hat{\Delta}_{l+1}, S_i - S_i \subset S - S \subset U - U \) and \( |S_i| \leq |S|-1 \leq r-1 \), thus \( \max_i F_A(S_i) \leq c_U(r - 1, l + 1) \) and (4.17) yields (4.16). Since \( c_U(1, l) = 0 \), (4.16) yields
\[
c_U(r,0) \leq C(d) \sum_{l=0}^{\infty} \left( \hat{\Delta}_l + 2^l \sqrt{p} \Delta_l + 2^{-l} \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A) \right).
\]
For \( U \subset (\frac{1}{2}(B_{d-1}^n)) \times T \), we have by (4.15)
\[
F_A(U) = \sup \{ F_A(S) : S \subset U, |S| < \infty \} \leq \sup r c_U(r,0)
\]
\[
\leq C(d) \left( \sqrt{p} \Delta_A(U) + \sum_{k=0}^{d-1} p^{\frac{k+d}{2}} s_k(A) \right).
\]
Finally, if $U \subset (B^n_2)^{d-1} \times T$, then $U' := \{(y/2, t) : (y, t) \in U\} \subset (B^n_2)^{d-1} \times T$ and

$$
\Delta_A(U') = 2^{d-1} \Delta_A(U),
$$

hence,

$$
F_A(U) = 2^{d-1} F_A(U') \leq C(d) \left( \sqrt{p} \Delta_A(U) + \sum_{k=0}^{d-1} p^{k+1-d} s_k(A) \right).
$$

\[ \square \]

5 Proofs of main results

We return to the notation used Section 2. In particular in this section the multi-index $i$ takes values in $[n]^d$ (instead of $[n]^d \times [m]$ as we had in the two previous sections) and all summations over $i$ should be understood as summations over $[n]^d$.

5.1 Proofs of Theorems 2.1 and 2.5

Proof of Theorem 2.1. We start with the lower bound. Fix $J \subset [d]$, $\mathcal{P} \in \mathcal{P}([d] \setminus J)$ and observe that

$$
\left\| \sum_i a_i \prod_{k=1}^d g_{ik}^{(k)} \right\|_p \geq \left( \mathbb{E}(G^{(i)} : j \in J) \sup_{\varphi \in F^*} \left| \varphi \left( \sum_i a_i \prod_{k=1}^d g_{ik}^{(k)} \right) \right| ^p \right)^{1/p}.
$$

$$
\geq c(d) \left( \mathbb{E}(G^{(j)} : j \in \mathcal{P} \cup \mathcal{P}' \cup \{p\}) \left( \sum_{i,j} a_i \prod_{j \in J} g_{ij}^{(j)} \right) \left( \sum_{i \notin [d] \setminus J} a_i \prod_{j \in J} g_{ij}^{(j)} \right) \right)^{1/p}.
$$

$$
\geq c(d)p^{\frac{p}{2}} \mathbb{E} \left( \sum_{i,j} a_i \prod_{j \in J} g_{ij}^{(j)} \right) = c(d)p^{\frac{p}{2}} \|A\|_\mathcal{P}.
$$

where $F^*$ is the dual space and in the second inequality we used Theorem A.6.

The upper bound will be proved by an induction on $d$. For $d = 2$ it is shown in [2]. Suppose that $d \geq 3$ and the estimate holds for $F$-valued matrices of order $2, \ldots, d - 1$. By the induction assumption, we have

$$
\left\| \sum_i a_i \prod_{k=1}^d g_{ik}^{(k)} \right\|_p \leq C(d) \sum_{(\mathcal{P}, \mathcal{P}') \in \mathcal{P}([d] \setminus J)} p^{\left\| \mathcal{P} \right\|_p} \left\| \sum_{i=1}^d a_i g_{id} \right\|_{\mathcal{P}' \setminus \mathcal{P}}.
$$

\[(5.1)\]

Since $\|\cdot\|_{\mathcal{P}' \setminus \mathcal{P}}$ is a norm Lemma A.5 yields

$$
\left\| \sum_{i=1}^d a_i g_{id} \right\|_{\mathcal{P}' \setminus \mathcal{P}} \leq CE \left\| \sum_{i=1}^d a_i g_{id} \right\|_{\mathcal{P}' \setminus \mathcal{P}} + C \sqrt{p} \|A\|_{\mathcal{P}' \setminus \mathcal{P}, [d]}.
$$

\[(5.2)\]

Choose $\mathcal{P} = (I_1, \ldots, I_k), \mathcal{P}' = (J_1, \ldots, J_m)$ and denote $J = \bigcup \mathcal{P}'$. By the definition of $\|A\|_{\mathcal{P}' \setminus \mathcal{P}}$ we have

$$
\left\| \sum_{i=1}^d a_i g_{id} \right\|_{\mathcal{P}' \setminus \mathcal{P}} = \sup \left\{ \mathbb{E}(G^{(1)}, \ldots, G^{(m)}) \left\| \sum_i a_i x_{i_1}^{(1)} \cdots x_{i_k}^{(k)} \prod_{l=1}^m g_{il}^{(l)} \right\|_{\mathcal{P}' \setminus \mathcal{P}}, \forall j=1, \ldots, k \sum_{i_{i_j}} (x_{i_{i_j}}) = 1 \right\}.
$$

$$
= \sup \left\{ \left\| \sum_i a_i x_{i_1}^{(1)} \cdots x_{i_k}^{(k)} g_{id} \right\|_{i_1, \ldots, i_k \notin J} \left\| \sum_{i_{i_j}} (x_{i_{i_j}}) = 1 \right\},
$$

\[(5.3)\]
where \( G^{(i)} = (g^{(i)}_{i,j})_{i,j} \) and \( \| \cdot \| \) is a norm on \( F^{n,J} \) given by
\[
\| (b_{i,j})_{i,j} \| = E \left| \sum_{i,j} b_{i,j} \prod_{i=1}^{m} g^{(i)}_{i,j} \right|.
\]
Theorem 2.6 implies that
\[
E \sup \left\{ \left\| \sum_{i,j} a_i x_{i,j}^{(1)} \cdots x_{i,j}^{(k)} g^{(d)}_{i,j} \right\| : \forall j=1 \ldots k \sum_{i,j} (x_{i,j}^{(j)})^2 = 1 \right\}
\leq C(k) \sum_{(R',R) \in \mathcal{P}(d) \setminus \mathcal{J}} p^{\frac{|R|-k}{2}} \| A \|_{R',R} = C(k) \sum_{(R',R) \in \mathcal{P}(d) \setminus \mathcal{J}} p^{\frac{|R|-k}{2}} \| A \|_{R',R},
\]
where \( A := (\hat{a}_{i,j})_{i,j} \) is \( F^{n,J} \)-valued matrix of order \( d - |J| \) given by \( \hat{a}_{i,j} = (a_i)_{i,j} \) and \( \| A \|_{R',R} \) is defined in a similar way as \( \| A \|_{R'} \) but under the expectation occurs the norm \( \| \cdot \| \).

The above and (5.3) yield
\[
E \left| \sum_{i=1}^k g_{i,d} \right|_{p'} \leq C(k) \sum_{(R',R) \in \mathcal{P}(d)} p^{\frac{|R|-k}{2}} \| A \|_{R',R}.
\]
(5.4)

Since \( |\mathcal{P}| = k \) the theorem follows from (5.1), (5.2) and (5.4).

**Proof of Theorem 2.5.** Let \( S = \left| \sum_{i=1}^k a_i g_{i,d} \right| \). Chebyshev’s inequality and Theorem 2.1 yield for \( p > 0 \),
\[
P \left( S \geq C(d) \sum_{(P,P') \in \mathcal{P}(d)} p^{\frac{|P|}{2}} \| A \|_{P'} \right) \leq e^{-p}.
\]
(5.5)

Now we substitute
\[
t = C(d) \sum_{P' \in \mathcal{P}(d)} \| A \|_{P'} + C(d) \sum_{|P| \geq 1} p^{\frac{|P|}{2}} \| A \|_{P'} := t_1 + t_2
\]
and observe that if \( t_1 < t_2 \) then
\[
p \geq \frac{1}{C(d)} \min_{|P'| \geq 0} \left( \frac{t}{\| A \|_{P'}} \right)^{2/|P'|}.
\]
The first inequality of the theorem follows then by adjusting the constants.

On the other hand by the Paley-Zygmund inequality we get for \( p \geq 2 \),
\[
P \left( S \geq C^{-1}(d) \sum_{J \in [d]} \sum_{P \in \mathcal{P}(J)} p^{\frac{|P|}{2}} \| A \|_{P} \right) \geq P \left( \sum_{P} p^{\frac{|P|}{2}} \| A \|_{P} \right) \geq \left( 1 - \frac{1}{2p} \right)^{2} \frac{(E S^p)^2}{E S^{2p}} \geq e^{-C(d)p},
\]
where in the last inequality we used Theorem A.1. The inequality follows by a similar substitution as for the upper bound.
5.2 Proof of Proposition 2.11 and Theorem 2.10

Let us first note that Proposition 2.11 reduces (2.8) of Theorem 2.10 to the lower estimate given in Theorem 2.1, while (2.10) is reduced to Corollary 2.9. The tail bounds (2.9) and (2.11) can be then obtained by Chebyshev’s and Paley-Zygmund inequalities as in the proof of Theorem 2.5. The rest of this section will be therefore devoted to the proof of Proposition 2.11.

The overall strategy of the proof is similar to the one used in [4] to obtain the real valued case of Theorem 2.10. It relies on a reduction of inequalities for general polynomials of degree $D$ to estimates for decoupled chaoses of degree $d = 1, \ldots, D$. To this end we will approximate general polynomials by tetrahedral ones and split the latter into homogeneous parts of different degrees, which can be decoupled. The splitting may at first appear crude but it turns out that up to constants depending on $D$ one can in fact invert the triangle inequality, which is formalized in the following result due to Kwapień (see [13, Lemma 2]). Recall that a multivariate polynomial is called tetrahedral, if it is affine in each variable.

**Theorem 5.1.** If $X = (X_1, \ldots, X_n)$ where $X_i$ are independent symmetric random variables, $Q$ is a multivariate tetrahedral polynomial of degree $D$ with coefficients in a Banach space $E$ and $Q_d$ is its homogeneous part of degree $d$, then for any symmetric convex function $\Phi: E \to \mathbb{R}_+$ and any $d \in \{0, 1, \ldots, D\}$,

$$E\Phi(Q_d(X)) \leq E\Phi(C_D Q(X)).$$

It will be convenient to have the polynomial $f$ represented as a combination of multivariate Hermite polynomials:

$$f(x_1, \ldots, x_n) = \sum_{d=0}^{D} \sum_{d \in \Delta_d^n} a_d h_d(x_1) \cdots h_d(x_n), \quad (5.6)$$

where

$$\Delta_d^n = \{d = (d_1, \ldots, d_n) : \forall k \in [n] d_k \geq 0 \text{ and } d_1 + \cdots + d_n = d\}$$

and $h_m(x) = (-1)^m x^2/2 \frac{d^m}{dx^m} e^{-x^2/2}$ is the $m$-th Hermite polynomial. Recall that Hermite polynomials are orthogonal with respect to the standard Gaussian measure, in particular if $g$ is a standard Gaussian variable, then for $m \geq 1$, $E h_m(g) = 0$ (we will use this property several times without explicitly referring to it).

In what follows, we will use the following notation. For a set $I$, by $I^k$ we will denote the set of all one-to-one sequences of length $k$ with values in $I$. For an $F$-valued $d$-indexed matrix $A = (a_{i_1, \ldots, i_d})_{i_1, \ldots, i_d \leq n}$ and $x \in \mathbb{R}^{n^2} \simeq (\mathbb{R}^n)^{\otimes d}$ we will denote

$$\langle A, x \rangle = \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} x_{i_1, \ldots, i_d}.$$

Let $(W_t)_{t \in [0,1]}$ be a standard Brownian motion. Consider standard Gaussian random variables $g = W_1$ and, for any positive integer $N$,

$$g_{j,N} = \sqrt{N} (W_{t_{j,N}} - W_{t_{j-1,N}}), \quad j = 1, \ldots, N.$$

For any $d \geq 0$, we have the following representation of $h_d(g) = h_d(W_1)$ as a multiple stochastic integral (see [12, Example 7.12 and Theorem 3.21]),

$$h_d(g) = d! \int_0^1 \cdots \int_0^{t_d} dW_{t_1} \cdots dW_{t_{d-1}} dW_{t_d}.$$
Approximating the multiple stochastic integral leads to

$$h_d(g) = d! \lim_{N \to \infty} N^{-d/2} \sum_{1 \leq j_1 < \cdots < j_d \leq N} g_{j_1,N} \cdots g_{j_d,N}$$

$$= \lim_{N \to \infty} N^{-d/2} \sum_{j \in [N]} g_{j,N}$$ \hspace{1cm} (5.7)

where the limit is in $L^2(\Omega)$ (see [12, Theorem 7.3. and formula (7.9)]) and actually the convergence holds in any $L^p$ (see [12, Theorem 3.50]).

Now, consider $n$ independent copies $(W^{(i)}_t)_{t \in [0,1]}$ of the Brownian motion ($1 \leq i \leq n$) together with the corresponding Gaussian random variables: $g^{(i)} = W^{(i)}_1$ and, for $N \geq 1$,

$$g^{(i)}_{j,N} = \sqrt{N}(W^{(i)}_t - W^{(i)}_{t-})_j, \quad j = 1, \ldots, N.$$

Let also

$$G^{(n,N)} = (g^{(1)}_{1,N}, \ldots, g^{(1)}_{n,N}, g^{(2)}_{1,N}, \ldots, g^{(2)}_{n,N}, \ldots, g^{(n)}_{1,N}, \ldots, g^{(n)}_{n,N}) = (g^{(i)}_{j,N})_{(i,j) \in [n] \times [N]}$$

be a Gaussian vector with $nN$ coordinates. We identify here the set $[nN]$ with $[n] \times [N]$ via the bijection $(i,j) \leftrightarrow (i-1)N + j$. We will also identify the sets $([n] \times [N])^d$ and $[n]^d \times [N]^d$ in a natural way. For $d \geq 0$ and $d \in \Delta^N_d$, let

$$I_d = \{ i \in [n]^d : \forall l \in [n] \#i^{-1}\{l\} = d_l \},$$

and define a $d$-indexed matrix $B^{(N)}_d$ of $n^d$ blocks each of size $N^d$ as follows: for $i \in [n]^d$ and $j \in [N]^d$,

$$(B^{(N)}_d)_{(i,j)} = \begin{cases} \frac{d_1 \cdots d_d}{d^d} N^{-d/2} & \text{if } i \in I_d \text{ and } (i,j) := ((i_1, j_1), \ldots, (i_d, j_d)) \in ([n] \times [N])^d, \\ 0 & \text{otherwise.} \end{cases}$$

**Proof of Proposition 2.11.** Assume that $f$ is of the form (5.6), By [4, Lemma 4.3], for any $p > 0$,

$$\langle B^{(N)}_d, (G^{(n,N)})^{\otimes d} \rangle \xrightarrow{N \to \infty} h_d, (g^{(1)}) \cdots h_d, (g^{(n)}) \text{ in } L^p(\Omega),$$

which together with the triangle inequality implies that

$$\lim_{N \to \infty} \left\| \sum_{d=1}^D \left( \sum_{d \in \Delta^N_d} a_d B^{(N)}_d, (G^{(n,N)})^{\otimes d} \right) \right\|_p = \| f(G) - Ef(G) \|_p$$

for any $p > 0$, where $G = (g^{(1)}, \ldots, g^{(n)})$ and we interpret multiplication of an element of $F$ and a real valued $d$ indexed matrix in a natural way. Thus, by Theorem 5.1 and the triangle inequality we obtain

$$C^{-1}_D \lim_{N \to \infty} \sum_{d=1}^D \left\| \sum_{d \in \Delta^N_d} a_d B^{(N)}_d, (G^{(n,N)})^{\otimes d} \right\|_p \leq \| f(G) - Ef(G) \|_p$$

$$\leq \lim_{N \to \infty} \sum_{d=1}^D \left\| \sum_{d \in \Delta^N_d} a_d B^{(N)}_d, (G^{(n,N)})^{\otimes d} \right\|_p$$

(recall that the matrices $B^{(N)}_d$ have zeros on generalized diagonals and so do their linear combinations).
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Denote by \( G^{(n,N,1)}, \ldots, G^{(n,N,d)} \) independent copies of \( G^{(n,N)} \).

By decoupling inequalities of Theorem A.9 we have

\[
\left\| \sum_{d \in \Delta_d^N} a_d B_d^{(N)}(g) \right\|_p = \left\| \sum_{d \in \Delta_d^N} a_d B_d^{(N)}(G^{(n,N,1)} \otimes \cdots \otimes G^{(n,N,d)}) \right\|_p. 
\] (5.8)

To finish the proof it is therefore enough to show that for any \( d \leq D \),

\[
\lim_{N \to \infty} \left\| \sum_{d \in \Delta_d^N} a_d B_d^{(N)}(G^{(n,N,1)} \otimes \cdots \otimes G^{(n,N,d)}) \right\|_p = \frac{1}{d!} \left\langle A_d, G_1 \otimes \cdots \otimes G_d \right\rangle_p, 
\] (5.9)

where \( G_1, \ldots, G_D \) are independent copies of \( G \) (recall that \( A_d = \mathbb{E} \nabla^d f(G) \)).

Fix \( d \geq 1 \). For any \( d \in \Delta_d^N \) define a symmetric \( d \)-indexed matrix \( \{b_d\}_{i \in [n]^d} \) as

\[
(b_d)_i = \begin{cases} 
\frac{d! d_1 \cdots d_n}{d!} & \text{if } i \in I_d, \\
0 & \text{otherwise},
\end{cases}
\]

and a symmetric \( d \)-indexed matrix \( \{\tilde{B}_d^{(N)}\}_{(i,j) \in ([n] \times [N])^d} \) as

\[
(\tilde{B}_d^{(N)})_{(i,j)} = N^{-d/2} (b_d)_i \text{ for all } i \in [n]^d \text{ and } j \in [N]^d.
\]

Using the convolution properties of Gaussian distributions one easily obtains

\[
\left\| \sum_{d \in \Delta_d^N} a_d \tilde{B}_d^{(N)}(G^{(n,N,1)} \otimes \cdots \otimes G^{(n,N,d)}) \right\|_p = \left\| \sum_{d \in \Delta_d^N} a_d (b_d)_{i \in [n]^d}, G_1 \otimes \cdots \otimes G_d \right\|_p. 
\] (5.10)

On the other hand, for any \( d \in \Delta_d^N \), the matrices \( \tilde{B}_d^{(N)} \) and \( B_d^{(N)} \) differ at no more than \( |I_d| \cdot (|[N]^d \setminus [N]^d)| \) entries. Thus

\[
\left\| a_d \left( \tilde{B}_d^{(N)} - B_d^{(N)}, G^{(n,N,1)} \otimes \cdots \otimes G^{(n,N,d)} \right) \right\|_p \\
\leq C(d) p^2 \|a_d\| \left\| \left( \tilde{B}_d^{(N)} - B_d^{(N)}, G^{(n,N,1)} \otimes \cdots \otimes G^{(n,N,d)} \right) \right\|_2 \\
\leq C(d) p^2 \|a_d\| \sqrt{|I_d|} \left( \frac{d! d_1 \cdots d_n}{d!} \right)^2 N^{-d} \left( \frac{N!}{(N-d)!} \right) \to 0
\]

as \( N \to \infty \), where in the first inequality we used Theorem A.1.

Together with the triangle inequality and (5.10) this gives

\[
\lim_{N \to \infty} \left\| \sum_{d \in \Delta_d^N} a_d B_d^{(N)}(G^{(n,N,1)} \otimes \cdots \otimes G^{(n,N,d)}) \right\|_p \\
= \left\| \sum_{d \in \Delta_d^N} a_d (b_d)_{i \in [n]^d}, G_1 \otimes \cdots \otimes G_d \right\|_p. 
\] (5.11)

Finally, we have

\[
A_d = \mathbb{E} \nabla^d f(G) = d! \sum_{d \in \Delta_d^N} a_d (b_d)_{i \in [n]^d}. 
\] (5.12)

Indeed, using the identity on Hermite polynomials, \( \frac{d}{dx} h_k(x) = k h_{k-1}(x) \) \( (k \geq 1) \), we obtain \( \mathbb{E} \frac{d}{dx} h_k(g) = k! \mathbf{1}_{k=1} \) for \( k, l \geq 0 \), and thus, for any \( d, l \leq D \) and \( d \in \Delta_d^N \),

\[
(\mathbb{E} \nabla^d h_d(g^{(1)} \cdots h_d(g^{(n)})))_i = d! (b_d)_i \mathbf{1}_{d=l} \text{ for each } i \in [n]^d.
\]

Now (5.12) follows by linearity. Combining it with (5.11) yields (5.9) and ends the proof. \(\square\)
5.3 Proof of the bound for chaoses in exponential variables

Proof of Proposition 2.18. Lemma A.8 implies

$$\left\| \sum_{i} a_i \prod_{k=1}^{d} E_{ik}^{(k)} \right\|_p \sim d \left\| \sum_{i_1, \ldots, i_{2d}} \hat{a}_{i_1, \ldots, i_{2d}} \prod_{k=1}^{2d} \hat{g}_{ik}^{(k)} \right\|_p, $$

(5.13)

where

$$\hat{a}_{i_1, \ldots, i_{2d}} := a_{i_1, \ldots, i_{d}} I_{\{i_1=i_{d+1}, \ldots, i_d=i_{2d}\}}.$$

Let \( \hat{A} = (\hat{a}_{i_1, \ldots, i_{2d}})_{i_1, \ldots, i_{2d}} \).

Theorem 2.15 and (5.13) yield

$$\frac{1}{C(d)} q^{1/2-d} \sum_{J \subset [2d]} \sum_{P \in P(J)} p^{\frac{|J|}{2}} \| \hat{A} \|_{L_q, P} \leq \left\| \sum_{i} a_i \prod_{k=1}^{d} E_{ik}^{(k)} \right\|_p \leq C(d) q^{2d-\frac{d}{2}} \sum_{J \subset [2d]} \sum_{P \in P(J)} p^{\frac{|J|}{2}} \| \hat{A} \|_{L_q, P}. $$

(5.14)

We will now express \( \sum_{J \subset [2d]} \sum_{P \in P(J)} p^{\frac{|J|}{2}} \| \hat{A} \|_{L_q, P} \) in terms of the matrix \( A \). To this end we need to introduce new notation. Consider a finite sequence \( M = (J, I_1, \ldots, I_k) \) of subsets of \( [d] \), such that \( J \cup I_1 \cup \ldots \cup I_k = [d], I_1, \ldots, I_k \neq \emptyset \) and each number \( m \in [d] \) belongs to at most two of the sets \( J, I_1, \ldots, I_k \). Denote the family of all such sequences by \( M([d]) \). For \( M = (J, I_1, \ldots, I_k) \) set \( |M| = k + 1 \) and

$$\langle A \rangle_{L_q, M} := \sup \left\{ \left\| \sum_{i \in J} \left( \sum_{r=1}^{k} x^{(r)}_{i_{|i|, j}} \right)^2 \right\|_{L_q} \right\},$$

where we do not exclude that \( J = \emptyset \). By a straightforward verification

$$\sum_{J \subset [2d]} \sum_{P \in P(J)} p^{\frac{|J|}{2}} \| \hat{A} \|_{L_q, P} \sim d \sum_{M \in M([d])} p^{\frac{|M|-1}{2}} \langle A \rangle_{L_q, M}. $$

(5.15)

To finish the proof it is enough to show that

$$\sum_{M \in M([d])} p^{\frac{|M|-1}{2}} \langle A \rangle_{L_q, M} \sim d \sum_{M \in C} p^{\frac{|M|-1}{2}} \langle A \rangle_{L_q, M}, $$

(5.16)

where

$$C = \left\{ M = (J, I_1, \ldots, I_k) \in M([d]) \mid J \cap \left( \bigcup_{l=1}^{k} I_l \right) = \emptyset, \right. \left. \forall l, m \leq k I_m \cap I_l \neq \emptyset \Rightarrow (|I_l| = |I_m| = 1, I_l = I_m) \right\}. $$

Indeed assume that (5.16) holds and choose \( M = (J, I_1, \ldots, I_k) \in C \). Consider
\[ I = \{ i \mid \exists_{l \leq m \leq k} \{ i \} = I_l = I_m \}. \] Then \( J \cap I = \emptyset \) and we have

\[
\langle A \rangle_{L^q,\mathcal{M}}^q = \sup \left\{ \int_V \left( \sum_{i,j} \left( \sum_{i_{1:} \in I} \frac{a_{i_1, \ldots, i_d}(v)}{\prod_{l \leq k} y_{i_l}^{(l)} x_{i_{l_j}}^{(l)}} \right)^2 \right)^{q/2} \right|_{\forall I \leq k} \left( \sum_{i_{1:} \in I} (x_{i_{1_j}}^{(l)})^2 \right) \leq 1, \forall I \in I \left( \bigcap_{l \in I} I_l \cap I = \emptyset \right) \}
\]

\[
= \max_{i_j} \left( \| (a_{i_{1:}, \ldots, i_d})_{i_{1:}} \|_{L_q,\{ I_l : I_l \cap I = \emptyset \}}^q \right) = \max_{i_j} \left( \| (a_{i_{1:}, \ldots, i_d})_{i_{1:}} \|_{L_q,\mathcal{M}}^q \right), \tag{5.17}
\]

where in the second equality we used the fact that

\[
(y_{i_l}^{(l)}, x_{i_{l_j}}^{(l)})_{i_l} \subset B^1 = \{ x \in \mathbb{R}^N : \sum_{i=1}^N |x_i| \leq 1 \}
\]
together with convexity and homogeneity of the norm

\[
\| (f_{i,j})_{i,j} \|_{L^q(x)} = \left( \int_X \left( \sum_{i,j} f_{i,j}^2 \right)^{q/2} \right)^{1/q}.
\]

By combining the above with (5.14)–(5.16) and comparing the exponents of \( p \) we conclude the assertion of the proposition.

The proof is completed by showing that

\[
\sum_{\mathcal{M} \in \mathcal{M}(\mathcal{C}[d])} p^{\frac{|\mathcal{M}| - 1}{2}} \langle A \rangle_{L^q,\mathcal{M}} \leq C(d) \sum_{\mathcal{M} \in \mathcal{C}} p^{\frac{|\mathcal{M}| - 1}{2}} \langle A \rangle_{L^q,\mathcal{M}}
\]

(the other estimate in (5.16) is trivial), which will be done in two steps. Let us fix \( \mathcal{M} = (J, I_1, \ldots, I_k) \in \mathcal{M}(\mathcal{C}[d]) \).

**Step 1.** Assume first that \( J \cap (\bigcup_{I_{l=1}^k I_l}) \neq \emptyset \). Without loss of generality we can assume that \( 1 \in J \cap I_1 \). Denote \( \tilde{I}_1 = I_1 \setminus \{ 1 \} \) and for any matrix \( (x_{i_{1_j}}^{(1)})_{i_1} \) such that \( \sum_{i_{1_j}} (x_{i_{1_j}}^{(1)})^2 \leq 1 \), set \( (b_{i_{1_j}})_{i_1} := \left( \sqrt{\sum_{i_{1_j} \setminus \{ 1 \}} (x_{i_{1_j}}^{(1)})^2} \right)_{i_1} \). Clearly,

\[
(\tilde{b}_{i_{1_j}})_{i_1} \subset B^1 \quad \text{and} \quad \sum_{i_{1_j} \setminus \{ 1 \}} \left( \frac{x_{i_{1_j}}^{(1)}}{b_{i_1}} \right)^2 \leq 1.
\]

Observe that for any \( f_1, \ldots, f_n \in L^q(X, d\mu) \) the function

\[
[0, +\infty)^n \ni t \to \int_V \left( \sum f_{i,j}^2(v) t_i \right)^{q/2} \ d\mu(v)
\]
is convex (recall that \( q \geq 2 \)). Since \( B^n_q \cap [0, \infty)^n = \text{conv}(0, e_1, e_2, \ldots, e_n) \), we have

\[
\langle A \rangle^q_{L^q, \mathcal{M}} = \sup \left\{ \int_V \left( \sum_{i,j} b_{i,j} \left( \sum_{i,j} a_{i_1, \ldots, i_d}(v) \frac{x_{i_1}^{(1)} \cdots x_{i_d}^{(k)}}{b_{i_1} \cdots b_{i_d}} \right)^q \right) d\mu(v) \left| \forall 1 \leq i \leq k \sum_{i,j} (x_{i,j}^{(l)})^2 \leq 1 \right. \right\}
\]

\[
\leq \max_{i_1} \sup_{i_1} \left\{ \int_V \left( \sum_{i,j} a_{i_1, \ldots, i_d}(v) y_{i_1} \prod_{l=2}^k x_{i_l}^{(l)} \right)^q d\mu(v) \left| \forall 1 \leq i \leq k \sum_{i,j} (x_{i,j}^{(l)})^2 \leq 1 \right. \right\}
\]

Let

\[
\mathcal{M}' = \begin{cases} (J \setminus \{1\}, \{1\}, \{1\}, I_1, \ldots, I_k) & \text{if } I_1 \neq \emptyset \\ (J \setminus \{1\}, \{1\}, \{1\}, I_2, \ldots, I_k) & \text{if } I_1 = \emptyset 
\end{cases}
\]

By the same argument as was used for the second equality in (5.17) we obtain that the right-hand side above equals \( \langle A \rangle_{L^q, \mathcal{M}'} \), which gives

\[
\langle A \rangle_{L^q, \mathcal{M}} \leq \langle A \rangle_{L^q, \mathcal{M}'}.
\]

Observe that

\[
p^{[\mathcal{M}'-1]/2} \langle A \rangle_{L^q, \mathcal{M}} \leq p^{[\mathcal{M}'-1]/2} \langle A \rangle_{L^q, \mathcal{M}'} \leq p^{[\mathcal{M}'-1]/2} \langle A \rangle_{L^q, \mathcal{M}'}.
\]

By iterating this argument we obtain that \( p^{[\mathcal{M}'-1]/2} \langle A \rangle_{L^q, \mathcal{M}} \leq p^{[\mathcal{M}'-1]/2} \langle A \rangle_{L^q, \mathcal{M}'} \) for some \( \mathcal{M}'' = (J'', I_1'', \ldots, I_m'') \) such that \( J'' \cap \bigcup_{i=1}^m I_i'' = \emptyset \).

**Step 2.** Assume that for some \( l, m \leq k I_l \cap I_m \neq \emptyset \) and \( |I_l| \geq 2 \) or \( |I_m| \geq 2 \).

Without loss of generality assume that \( 1 \in I_1 \cap I_2 \) and \( |I_1| \geq 2 \). Clearly,

\[
\langle A \rangle^q_{L^q, \mathcal{M}} = \sup \left\{ \int_V \left( \sum_{i,j} \left( \sum_{i,j} a_{i_1, \ldots, i_d}(v) b_{i_1} c_{i_1} \frac{x_{i_1}^{(1)} \cdots x_{i_d}^{(k)}}{b_{i_1} c_{i_1}} \right)^q \right) d\mu(v) \left| \forall 1 \leq i \leq k \sum_{i,j} (x_{i,j}^{(l)})^2 \leq 1 \right. \right\},
\]

where \( (b_{i_1}, c_{i_1})_i := (\sqrt{\sum_{i_1, i_1 \setminus \{1\}} (x_{i_1}^{(1)})^2}_i, (c_{i_1})_i := (\sqrt{\sum_{i_2, i_2 \setminus \{1\}} (x_{i_2}^{(2)})^2}_i) \in B^n_q. \) Because \( (b_{i_1}, c_{i_1})_i \in B^n_q \),

\[
\forall i_1 \sum_{i_1, i_1 \setminus \{1\}} \left( \frac{x_{i_1}^{(1)}}{b_{i_1}} \right)^2 \leq 1, \quad \sum_{i_2, i_2 \setminus \{1\}} \left( \frac{x_{i_2}^{(2)}}{c_{i_1}} \right)^2 \leq 1.
\]
and for any \((f_{ij})_{ij}\) in \(L^q(X, d\mu)\), the function
\[
R^n \ni t \mapsto \int_V \left( \sum_{i} \left( \sum_{j} t_{ij} f_{ij}(v) \right) \right)^{q/2} d\mu(v)
\]
is convex, we obtain similarly as in Step 1,
\[
p^{\lvert M \rvert - 1/2} \langle A \rangle_{L_q,M} \leq p^{\lvert M \rvert - 1/2} \langle A \rangle_{L_q,M'} \leq p^{\lvert M \rvert - 1/2} \langle A \rangle_{L_q,M'}
\]
where
\[
M' := \begin{cases} (J, \{1\}, I_1 \setminus \{1\}, I_2 \setminus \{1\}, I_3, \ldots, I_k) & \text{if } I_2 \setminus \{1\} \neq \emptyset \\ (J, \{1\}, I_1 \setminus \{1\}, I_3, \ldots, I_k) & \text{otherwise} \end{cases}
\]
An iteration of this argument shows that indeed one can assume that \(M\) satisfies the implication \(I_m \cap I_t \neq \emptyset \Rightarrow (\lvert I_t \rvert = \lvert I_m \rvert = 1, I_t = I_m)\).

Combining Steps 1 and 2 we obtain that for any \(M \in M(d)\) there exists \(M' \in \mathcal{C}\) such that \(p^{\lvert M \rvert - 1/2} \langle A \rangle_{L_q,M} \leq p^{\lvert M \rvert - 1/2} \langle A \rangle_{L_q,M'}\) which yields (5.16). \(\Box\)

A Appendix

In this section we gather technical facts that are used in the proof.

**Theorem A.1** (Hypercontractivity of Gaussian chaoses). Let
\[
S = a + \sum_{i_1} a_{i_1} g_{i_1} + \sum_{i_1, i_2} a_{i_1, i_2} g_{i_1} g_{i_2} + \cdots + \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1} \cdots g_{i_d},
\]
be a non-homogeneous Gaussian chaos of order \(d\) with values in a Banach space \((F, \|\cdot\|)\). Then for any \(1 \leq p < q < \infty\), we have
\[
\left( \mathbb{E} \| S \|^q \right)^{1/q} \leq C(d) \left( \frac{q}{p} \right)^{d/2} \left( \mathbb{E} \| S \|^p \right)^{1/p}.
\]

**Proof.** It is an immediate consequence of [9, Theorem 3.2.10] and Hölder’s inequality. \(\Box\)

**Theorem A.2** (Sudakov minoration [27]). For any set \(T \subset R^n\) and \(\varepsilon > 0\) we have
\[
\varepsilon \sqrt{\ln N(T, d_2, \varepsilon)} \leq C \varepsilon \sup_{t \in T} \sum_i t_i g_i,
\]
where \(d_2\) is the Euclidean distance.

**Theorem A.3** (Dual Sudakov minoration [17, formula (3.15)]). Let \(\alpha\) be a norm on \(R^n\) and \(\rho_n(x, y) = \alpha(x - y)\) for \(x, y \in R^n\). Then
\[
\varepsilon \sqrt{\ln N(B^2_n, \rho_n, \varepsilon)} \leq C \varepsilon \alpha(G_n) \quad \text{for } \varepsilon > 0.
\]

**Lemma A.4** ([15, Lemma 3]). Let \((G_t)_{t \in T}\) be a centered Gaussian process and \(T = \bigcup_{i=1}^m T_i, m \geq 1\). Then
\[
\mathbb{E} \sup_{t \in T} G_t \leq \max_{\ell \leq m} \mathbb{E} \sup_{t \in T} G_t + C \sqrt{\ln(m)} \sup_{t, t' \in T} \sqrt{\mathbb{E}(G_t - G_{t'})^2}.
\]

**Lemma A.5.** Let \(G\) be a Gaussian variable in a Banach space \((F, \|\cdot\|)\). Then for any \(p \geq 2\),
\[
\frac{1}{C} \left( \|G\|_1 + \sqrt{p} \sup_{\varphi \in F^*} \mathbb{E} |\varphi(G)| \right) \leq \|G\|_p \leq \|G\|_1 + C \sqrt{p} \sup_{\varphi \in F^*, \|\varphi\| \leq 1} \mathbb{E} |\varphi(G)|,
\]
where \((F^*, \|\cdot\|_*)\) is the dual space to \((F, \|\cdot\|)\).
Moments of Gaussian chaos in Banach spaces

In particular for \(a_1, \ldots, a_n \in F\) and \(G = \sum_{i=1}^n a_i g_i\) we have
\[
\left\| \sum_{i=1}^n a_i g_i \right\|_p \sim \mathbb{E} \left\| \sum_{i=1}^n a_i g_i \right\| + \sqrt{p} \sup_{\|x\|_2^2} \left\| \sum_{i=1}^n a_i x_i \right\|.
\]

Proof. The first part of the Lemma is [15, Lemma 4]. It implies the second part as follows. Since \(\varphi(G)\) is a one-dimensional Gaussian vector, we have \(\|\varphi(G)\|_2 \sim \mathbb{E} |\varphi(G)|\). In particular for \(G = \sum_{i=1}^n a_i g_i\), \(a_i \in F\) we obtain
\[
\sup_{\varphi \in F^*, \|\varphi\| \leq 1} \mathbb{E}|\varphi(G)| \sim \sup_{\varphi \in F^*, \|\varphi\| \leq 1} \|\varphi(G)\|_2 = \sup_{\varphi \in F^*, \|\varphi\| \leq 1} \left( \sum_i \varphi(a_i)^2 \right)^{1/2} = \sup_{\varphi \in F^*, \|\varphi\| \leq 1} \sup_i \sum_i \varphi(a_i) x_i = \sup_{x \in \ell^2} \left\| \sum_i a_i x_i \right\|. \quad \Box
\]

Theorem A.6 ([15, Theorem 1]). For any real-valued matrix \((a_{i_1, \ldots, i_d})_{i_1, \ldots, i_d}\) and \(p \geq 2\), we have
\[
\left\| \sum_{i} a_i \prod_{j=1}^d g_{i,j}^{(j)} \right\|_p \sim \sum_{\rho \in \Pi(d)} p^{\rho(j)/2} \sup \left\{ \sum_{i} a_i \prod_{j=1}^d x_{i,j}^{(j)} \mid \forall 1 \leq j \leq k, \left\| (x_{i,j})_{i,j} \right\|_2 \leq 1 \right\}.
\]

Corollary A.7. Assume that for any \(i_1, \ldots, i_d, a_{i_1, \ldots, i_d} \in \mathbb{R}\). Then for all \(p \geq 1\)
\[
\frac{1}{C(d)} \sqrt[p]{\sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d}^2} \leq \left\| \sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d} g_{i_1}^{(1)} \cdots g_{i_d}^{(d)} \right\|_p \leq C(d) \sqrt[p]{\sum_{i_1, \ldots, i_d} a_{i_1, \ldots, i_d}^2}.
\]

Proof. It is an easy consequence of Theorems A.1 and A.6. \(\Box\)

Lemma A.8 ([3, Lemma 9.5]). Let \(Y_i^{(1)}\) be independent standard symmetric exponential variables (variables with density \(2^{-1} \exp(-|t|))\) and \(Y_i^{(2)} = g_i^2\), \(Y_i^{(3)} = g_i g_0\), where \(g_i, g_0\) are i.i.d. \(N(0, 1)\) variables and \(\varepsilon_i\) i.i.d. Rademacher variables independent of \((Y_i^{(1)}), (Y_i^{(2)}), (Y_i^{(3)})\). Then for any Banach space \((F, \|\|)\), any \(p \geq 1\) and any vectors \(v_1, \ldots, v_n \in F\) the quantities
\[
\left( \mathbb{E} \left\| \sum_i v_i \varepsilon_i Y_i^{(j)} \right\|^p \right)^{1/p}, \quad j = 1, 2, 3,
\]
are comparable up to universal multiplicative factors.

We remark that the above lemma is formulated in [3] for \(p = 1\) in the real valued case, however the proof presented there (based on the contraction principle) works for arbitrary \(p \geq 1\) and arbitrary Banach spaces.

We will also need decoupling inequalities for tetrahedral homogeneous polynomials. Such inequalities were introduced for the first time in [18] for real valued multi-linear forms and since then have been strengthened and generalized by many authors (see the monograph [9]). The following theorem is a special case of results from [13] (treating also general tetrahedral polynomials) and [8, 10] (treating general U-statistics).

Theorem A.9. Let \(X = (X_1, \ldots, X_n)\) be a sequence of independent random variables and let \(X^{(l)} = (X_1^{(l)}, \ldots, X_n^{(l)})\), \(l = 1, \ldots, d\), be i.i.d. copies of \(X\). Consider a \(d\)-indexed symmetric matrix \((a_{i_1, \ldots, i_d})_{i_1, \ldots, i_d=1}^n\), with coefficients from a Banach space \(F\). Assume
that $a_{1,\ldots,i_d} = 0$ whenever there exist $1 \leq k < m \leq d$ such that $i_k = i_m$. Then for any $p \geq 1$,
\[
\left\| \sum_{i_1,\ldots,i_d=1}^{n} a_{i_1,\ldots,i_d} X_{i_1} \cdots X_{i_d} \right\|_p \sim d \left( \sum_{i_1,\ldots,i_d=1}^{n} a_{i_1,\ldots,i_d} X_{i_1}^{(1)} \cdots X_{i_d}^{(d)} \right)^{\frac{1}{p}}.
\]
Moreover, for any $t > 0$,
\[
C_d^{-1} P\left( \left\| \sum_{i_1,\ldots,i_d=1}^{n} a_{i_1,\ldots,i_d} X_{i_1}^{(1)} \cdots X_{i_d}^{(d)} \right\|_p \geq Ct \right) \leq P\left( \left\| \sum_{i_1,\ldots,i_d=1}^{n} a_{i_1,\ldots,i_d} X_{i_1} \cdots X_{i_d} \right\|_p \geq t \right) \leq C_d P\left( \left\| \sum_{i_1,\ldots,i_d=1}^{n} a_{i_1,\ldots,i_d} X_{i_1}^{(1)} \cdots X_{i_d}^{(d)} \right\|_p \geq t/C_d \right).
\]

B Glossary

- $F$ – the underlying Banach space, p. 2
- $(g_{i,k}^{(k)})_{i,k \geq 1}$ – an array of i.i.d. standard Gaussian variables, p. 2
- $F^*$ – the dual of $F$, p. 12
- $B_2^d$ – the standard unit Euclidean ball, p. 2
- $B_1^d$ – the standard unit ball in $\mathbb{R}^n$ in the norm $\| \cdot \|_1$, p. 30
- $[n]$ – the set $\{1, \ldots, n\}$, p. 5
- $\mathbb{R}^{q,d} = (\mathbb{R}^n)^{\otimes d}$ – the space of $d$-indexed matrices, p. 5
- $B_{1,d}$ – the unit Euclidean (Hilbert-Schmidt) ball in $\mathbb{R}^{n,d}$, p. 5
- $\mathcal{P}(I)$ – the set of partitions of $I$, p. 5
- $\| \cdot \|_p$ – a norm appearing in general moment estimates, p. 5
- $\| A \|_{L_q,p}$ – a norm appearing in general moment estimates, p. 5
- $G^{(i)} = (g_{i_1}^{(l)})_{i_1}$ – independent arrays of i.i.d. standard Gaussian variables, p. 5
- $\| A \|_{L_q,p}$ – a norm appearing in moment estimates for $L_q$-values chaoses, p. 10
- $s_k(A)$ – the sum of norms corresponding to partitions of cardinality $k$, p. 12
- $d_{\rho_A}$ – the distance induced by the canonical Gaussian process, p. 13
- $\alpha_A$ – a norm related to $\rho_A$, p. 13
- $\tilde{N}(S, \rho, \varepsilon)$ – the entropy number, p. 13
- $W_d^{s}(\alpha_A, s)$ – a quantity appearing in estimates for entropy numbers, p. 13
- $W_d^{s}(\alpha_A, s)$ – a quantity appearing in estimates for entropy numbers, p. 13
- $H_A$ – an auxiliary norm related to the main Gaussian process, p. 13
- $V_d^{s}(\beta_A)$ – a quantity appearing in estimates for entropy numbers, p. 13
- $\rho_{\beta_A}$ – a measure used in the volumetric argument, p. 14
- $d_A$ – an auxiliary norm appearing in estimates for entropy numbers, p. 16
- $F_A(U)$ – the expected supremum of the main Gaussian process, p. 17
- $d_A(U)$ – the diameter of the set $U$ in the distance $\rho_A$, p. 17
- $d_{\rho_A}$ – the set of multi-indices corresponding to Hermite polynomials of degree $d$, p. 26
- $(A)_{L,q,M}$ – an auxiliary norm related to moment estimates for chaoses in exponential variables, p. 29
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