

Bounds for the average L^p -extreme and the L^∞ -extreme discrepancy

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Abstract

The extreme or unanchored discrepancy is the geometric discrepancy of point sets in the d -dimensional unit cube with respect to the set system of axis-parallel boxes. For $2 \leq p < \infty$ we provide upper bounds for the average L^p -extreme discrepancy. With these bounds we are able to derive upper bounds for the inverse of the L^∞ -extreme discrepancy with optimal dependence on the dimension d and explicitly given constants.

1 Introduction

Let $\mathcal{R}_d$ be the set of all half-open axis-parallel boxes in the d -dimensional unit ball with respect to the maximum norm, i.e.,

$$\mathcal{R}_d = \{[x, y) \mid x, y \in [-1, 1]^d, x \leq y\},$$

where $[x, y) := [x_1, y_1) \times \dots \times [x_d, y_d)$ and inequalities between vectors are meant component-wise. It is convenient to identify $\mathcal{R}_d$ with

$$\Omega := \{(\underline{x}, \bar{x}) \in \mathbb{R}^{2d} \mid -\mathbf{1} \leq \underline{x} \leq \bar{x} \leq \mathbf{1}\},$$

where for any real scalar a we put $\mathbf{a} := (a, \dots, a) \in \mathbb{R}^d$. The L^p -extreme discrepancy of a point set $\{t_1, \dots, t_n\} \subset [-1, 1]^d$ is given by

$$D_p(t_1, \dots, t_n) := \left(\int_{\Omega} \left| \prod_{l=1}^d (\bar{x}_l - \underline{x}_l) - \frac{1}{n} \sum_{i=1}^n 1_{[\underline{x}, \bar{x})}(t_i) \right|^p d\omega(\underline{x}, \bar{x}) \right)^{1/p},$$

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where $1_{[\underline{x}, \overline{x})}$ denotes the characteristic function of $[\underline{x}, \overline{x})$ and $d\omega$ is the normalized Lebesgue measure $2^{-d} d\underline{x} d\overline{x}$ on Ω . The L^∞ -extreme discrepancy is

$$D_\infty(t_1, \dots, t_n) := \sup_{(\underline{x}, \overline{x}) \in \Omega} \left| \prod_{l=1}^d (\overline{x}_l - \underline{x}_l) - \frac{1}{n} \sum_{i=1}^n 1_{[\underline{x}, \overline{x})}(t_i) \right|,$$

and the smallest possible L^∞ -extreme discrepancy of any n -point set is

$$D_\infty(n, d) = \inf_{t_1, \dots, t_n \in [-1, 1]^d} D_\infty(t_1, \dots, t_n).$$

Another quantity of interest is the *inverse* of $D_\infty(n, d)$, namely

$$n_\infty(\varepsilon, d) = \min\{n \in \mathbb{N} \mid D_\infty(n, d) \leq \varepsilon\}.$$

If we consider in the definitions above the set of all d -dimensional corners

$$\mathcal{C}_d = \{[-\mathbf{1}, y) \mid y \in [-1, 1]^d\}$$

instead of $\mathcal{R}_d$, we get the classical notion of star-discrepancy.

It is well known that the star-discrepancy is related to the error of multivariate integration of certain function classes (see, e.g., [2, 5, 8, 10, 12]). That this is also true for the extreme discrepancy was pointed out by Novak and Woźniakowski in [12]. Therefore it is of interest to derive upper bounds for the extreme discrepancy with a good dependence on the dimension d and explicitly known constants.

Heinrich, Novak, Wasilkowski and Woźniakowski showed in [4] with probabilistic methods that for the inverse $n_\infty^*(\varepsilon, d)$ of the star-discrepancy we have $n_\infty^*(\varepsilon, d) \leq Cd\varepsilon^{-2}$. The drawback is here that the constant C is not known. In the same paper a lower bound was proved establishing the linear dependence of $n_\infty^*(\varepsilon, d)$ on d . This bound has recently been improved by Aicke Hinrichs to $n_\infty^*(\varepsilon, d) \geq cd\varepsilon^{-1}$ [6]. These results hold also for $n_\infty(\varepsilon, d)$.

In [4], Heinrich et al. presented two additional bounds for $n_\infty^*(\varepsilon, d)$ with slightly worse dependence on d , but explicitly known constants. The first one uses again a probabilistic approach, employs Hoeffding's inequality and leads to

$$n_\infty^*(\varepsilon, d) \leq O(d\varepsilon^{-2}(\ln(d) + \ln(\varepsilon^{-1}))).$$

The approach has been modified in more recent papers to improve this bound or to derive similar results in different settings [1, 5, 9]. In particular, it has been implicitly shown in the quite general Theorem 3.1 in [9] that the last bound holds also for the extreme discrepancy (as pointed out in [3], this result can be improved by employing the methods used in [1]).

The second bound was shown in the following way: The authors proved for even p an upper bound for the average L^p -star discrepancy $\text{av}_p^*(n, d)$:

$$\text{av}_p^*(n, d) \leq 3^{2/3} 2^{5/2+d/p} p(p+2)^{-d/p} n^{-1/2}.$$

(This analysis is quite elaborate, since $\text{av}_p^*(n, d)$ is represented as an alternating sum of weighted products of Stirling numbers of the first and second kind.) The bound was used to derive upper bounds $n_\infty^*(\varepsilon, d) \leq C_k d^2 \varepsilon^{-2-1/k}$ for every $k \in \mathbb{N}$. To improve the dependence on d , Hinrichs suggested to use symmetrization. This approach was sketched in [11] and leads to

$$\text{av}_p^*(n, d) \leq 2^{1/2+d/p} p^{1/2} (p+2)^{-d/p} n^{-1/2}$$

and $n_\infty^*(\varepsilon, d) \leq C_k d \varepsilon^{-2-1/k}$. (Actually there seems to be an error in the calculations in [11], therefore we stated the results of our own calculations—see Remark 4 and 9).

In this paper we use the symmetrization approach to prove an upper bound for the average L^p -extreme discrepancy $\text{av}_p(n, d)$ for $2 \leq p < \infty$. Our analysis does not need Stirling numbers and uses rather simple combinatorial arguments. Similar as in [4], we derive from this bound upper bounds for the inverse of the L^∞ -extreme discrepancy of the form $n_\infty(\varepsilon, d) \leq C_k d \varepsilon^{-2-1/k}$ for all $k \in \mathbb{N}$.

2 Bound for the average L^p -discrepancy

If $\underline{x}, \bar{x}$ are vectors in $\mathbb{R}^d$ with $\underline{x} \leq \bar{x}$, we use the (non-standard) notation $x := (\bar{x} - \underline{x})/2$. Let $p \in \mathbb{N}$ be even. For $i = 1, \dots, n$ we define the Banach space valued random variable $X_i : [-1, 1]^{nd} \rightarrow L^p(\Omega, d\omega)$ by $X_i(t)(\underline{x}, \bar{x}) = 1_{[\underline{x}, \bar{x}]}(t_i)$. Then $X_1, \dots, X_n$ are independent and identically distributed. Note that X_i is Bochner integrable for all $i \in [n]$. If $\mathbb{E}$ denotes the expectation with respect to the normalized measure $2^{-nd} dt$, then $\mathbb{E}X_i \in L^p(\Omega, d\omega)$ and $\mathbb{E}X_i(\underline{x}, \bar{x}) = x_1 \dots x_d$ almost everywhere. We obtain

$$\begin{aligned} \text{av}_p(n, d)^p &= \int_{[-1, 1]^{nd}} D_p(t_1, \dots, t_n)^p 2^{-nd} dt \\ &= \int_{[-1, 1]^{nd}} \left\| \frac{1}{n} \sum_{i=1}^n (X_i(t) - \mathbb{E}X_i) \right\|_{L^p(\Omega, d\omega)}^p 2^{-nd} dt \\ &= \mathbb{E} \left(\left\| \frac{1}{n} \sum_{i=1}^n (X_i - \mathbb{E}X_i) \right\|_{L^p(\Omega, d\omega)}^p \right). \end{aligned}$$

Let $\varepsilon_1, \dots, \varepsilon_n : [-1, 1]^{nd} \rightarrow \{-1, +1\}$ be symmetric Rademacher random variables, i.e., random variables taking the values ± 1 with probability $1/2$. We choose these variables such that $\varepsilon_1, \dots, \varepsilon_n, X_1, \dots, X_n$ are independent. Then (see [7, §6.1])

$$\begin{aligned} \text{av}_p(n, d)^p &\leq \mathbb{E} \left(2^p \left\| \frac{1}{n} \sum_{i=1}^n \varepsilon_i X_i \right\|_{L^p(\Omega, d\omega)}^p \right) \\ &= \left(\frac{2}{n} \right)^p \sum_{i_1, \dots, i_p=1}^n \int_{[-1, 1]^{nd}} \int_{\Omega} \prod_{l=1}^p \left(\varepsilon_{i_l}(t) 1_{[\underline{x}, \bar{x}]}(t_{i_l}) \right) d\omega(\underline{x}, \bar{x}) 2^{-nd} dt. \end{aligned}$$

Let us now consider $(\underline{x}, \bar{x}) \in \Omega$, $k \in [p]$, pairwise disjoint indices $i_1, \dots, i_k$ and $j_1, \dots, j_k \in [p]$ with $\sum_{l=1}^k j_l = p$. According to Fubini's Theorem

$$\begin{aligned} J &:= \int_{[-1,1]^{nd}} \left(\prod_{l=1}^k \varepsilon_{i_l}^{j_l}(t) \right) \left(\int_{\Omega} \left(\prod_{l=1}^k X_{i_l}^{j_l}(t)(\underline{x}, \bar{x}) \right) d\omega(\underline{x}, \bar{x}) \right) 2^{-nd} dt \\ &= \left(\prod_{l=1}^k \int_{[-1,1]^{nd}} \varepsilon_{i_l}^{j_l}(t) 2^{-nd} dt \right) \left(\int_{\Omega} \int_{[-1,1]^{nd}} \left(\prod_{l=1}^k 1_{[\underline{x}, \bar{x}]}(t_{i_l}) \right) 2^{-nd} dt d\omega(\underline{x}, \bar{x}) \right). \end{aligned}$$

This yields $J = \int_{\Omega} (x_1 \dots x_d)^k d\omega(\underline{x}, \bar{x}) = 2^d (k+1)^{-d} (k+2)^{-d}$ if every exponent j_l is even, and $J = 0$ if there exists at least one odd exponent j_l . Let $T(p, k, n)$ be the number of tuples $(i_1, \dots, i_p) \in [n]^p$ with $|\{i_1, \dots, i_p\}| = k$ and $|\{l \in [p] \mid i_l = i_m\}|$ even for each $m \in [p]$. Our last observation implies

$$\text{av}_p(n, d)^p \leq 2^{p+d} n^{-p} \sum_{k=1}^{p/2} \frac{T(p, k, n)}{(k+1)^d (k+2)^d}.$$

In the next step we shall estimate the numbers $T(p, k, n)$. For that purpose we introduce further notation. Let

$$M(p/2, k) = \left\{ \nu \in \mathbb{N}^k \mid 1 \leq \nu_1 \leq \dots \leq \nu_k \leq p/2, \sum_{i=1}^k \nu_i = p/2 \right\},$$

and for $\nu \in M(p/2, k)$ let $e(\nu, i) = |\{j \in [k] \mid \nu_j = i\}|$. With the standard notation for multinomial coefficients we get

$$T(p, k, n) = \sum_{\nu \in M(p/2, k)} \binom{p}{2\nu_1, \dots, 2\nu_k} \frac{n(n-1)\dots(n-k+1)}{e(\nu, 1)! \dots e(\nu, p/2)!}.$$

If $\sharp(p/2, k, n)$ denotes the number of tuples $(i_1, \dots, i_{p/2}) \in [n]^{p/2}$ with $|\{i_1, \dots, i_{p/2}\}| = k$, then

$$\sharp(p/2, k, n) = \sum_{\nu \in M(p/2, k)} \binom{p/2}{\nu_1, \dots, \nu_k} \frac{n(n-1)\dots(n-k+1)}{e(\nu, 1)! \dots e(\nu, p/2)!}.$$

We want to compare $T(p, k, n)$ with $\sharp(p/2, k, n)$ and are therefore interested in the quantity

$$Q_k^p(\nu) := \binom{p}{2\nu_1, \dots, 2\nu_k} \binom{p/2}{\nu_1, \dots, \nu_k}^{-1}.$$

To derive an upper bound for $Q_k^p(\nu)$, we prove two auxiliary lemmas.

Lemma 1. *Let $f : \mathbb{N}_0 \rightarrow \mathbb{R}$ be defined by $f(r) = [2r(2r-1)\dots(r+1)](2r)^{-r}$ for $r > 0$ and $f(0) = 1$. Then $f(r+s) \leq f(r)f(s)$ for all $r, s \in \mathbb{N}_0$.*

Proof. We prove the inequality for an arbitrary s by induction over r . It is evident if $r = 0$. So let the inequality hold for some $r \in \mathbb{N}_0$. The well known relations $\Gamma(x+1) = x\Gamma(x)$ and $\sqrt{\pi}\Gamma(2x) = 2^{2x-1}\Gamma(x)\Gamma(x+1/2)$ for the gamma function lead to

$$f(r) = 2^{2r} \pi^{-1/2} \Gamma(r+1/2) \exp(-r \ln(2r)),$$

with the convention $0 \cdot \ln(0) = 0$ when $r = 0$, and

$$\frac{f(r+1+s)}{f(r+1)} = \frac{g(r)}{g(r+s)} \frac{f(r+s)}{f(r)},$$

where $g : [0, \infty) \rightarrow [0, \infty)$ is defined by $g(0) = 2$ and

$$g(\lambda) = \left(1 + \frac{1}{2\lambda+1}\right) \exp(\lambda \ln(1 + 1/\lambda))$$

for $\lambda > 0$. The function g is continuous in 0 and its derivative is given by

$$g'(\lambda) = \left(\ln(1 + 1/\lambda) - \frac{2}{2\lambda+1}\right) g(\lambda)$$

for $\lambda > 0$. Since

$$\frac{d}{d\lambda} \left(\ln(1 + 1/\lambda) - \frac{2}{2\lambda+1} \right) = -\frac{1}{\lambda(\lambda+1)(2\lambda+1)^2} < 0$$

and

$$\lim_{\lambda \rightarrow \infty} \left(\ln(1 + 1/\lambda) - \frac{2}{2\lambda+1} \right) = 0,$$

we obtain $g'(\lambda) \geq 0$. Therefore g is an increasing function. Thus

$$\frac{g(r)}{g(r+s)} \leq 1, \text{ which establishes } \frac{f(r+1+s)}{f(r+1)} \leq \frac{f(r+s)}{f(r)} \leq f(s).$$

□

Lemma 2. Let $k \in \mathbb{N}$, $a_1, \dots, a_k \in [0, \infty)$ and $\sigma = \sum_{i=1}^k a_i$. Then

$$\left(\frac{\sigma}{k}\right)^\sigma \leq \prod_{i=1}^k a_i^{a_i}.$$

Proof. Let $\sigma > 0$, and consider the functions $s : \mathbb{R}^k \rightarrow \mathbb{R}$, $x \mapsto \sum_{i=1}^k x_i$ and

$$f : [0, \infty)^k \rightarrow \mathbb{R}, x \mapsto \prod_{i=1}^k x_i^{x_i} = \prod_{i=1}^k \exp(x_i \ln(x_i)),$$

where we use the convention $0 \cdot \ln(0) = 0$. Let $M = \{x \in (0, \infty)^k \mid s(x) = \sigma\}$. Since f is continuous, there exists a point ξ in the closure $\overline{M}$ of M with $f(\xi) = \min\{f(x) \mid x \in \overline{M}\}$.

Now let $x \in \overline{M} \setminus M$, which implies $x_\mu = 0$ for an index $\mu \in [k]$. Since $s(x) = \sigma$, there exists a $\nu \in [k]$ with $x_\nu > 0$. Without loss of generality we may assume $\mu = 1$, $\nu = 2$. Then

$$f(x) = \prod_{i=2}^k x_i^{x_i} > \left(\frac{x_2}{2}\right)^{x_2} \prod_{i=3}^k x_i^{x_i} = f(x'),$$

where $x' = (\frac{x_2}{2}, \frac{x_2}{2}, x_3, \dots, x_k)$. Thus ξ lies in M . Since $\text{grad } s \equiv (1, \dots, 1) \neq 0$, there exists a Lagrangian multiplier $\lambda \in \mathbb{R}$ with $\text{grad } f(\xi) = \lambda \text{grad } s(\xi)$. From $\text{grad } f(x) = (1 + \ln(x_1), \dots, 1 + \ln(x_k))f(x)$ follows $\xi_1 = \dots = \xi_k$, i.e., $\xi_i = \sigma/k$ for $i = 1, \dots, k$. $\square$

With the help of Lemma 1 and 2 we conclude

$$Q_k \leq \frac{p^{p/2}}{(2\nu_1)^{\nu_1} \dots (2\nu_k)^{\nu_k}} = \left(\prod_{i=1}^k \nu_i^{\nu_i}\right)^{-1} \left(\frac{p}{2}\right)^{p/2} \leq \left(\frac{1}{k} \sum_{i=1}^k \nu_i\right)^{-p/2} \left(\frac{p}{2}\right)^{p/2} = k^{p/2}.$$

Therefore

$$T(p, k, n) \leq k^{p/2} \#(p/2, k, n). \quad (1)$$

The last estimate yields

$$\text{av}_p(n, d)^p \leq 2^{p+d} n^{-p} \sum_{k=1}^{p/2} \frac{k^{p/2}}{(k+1)^d (k+2)^d} \#(p/2, k, n).$$

If $p \geq 4d$, then

$$\begin{aligned} \text{av}_p(n, d)^p &\leq 2^{p/2+3d} n^{-p} p^{p/2} (p+2)^{-d} (p+4)^{-d} \sum_{k=1}^{p/2} \#(p/2, k, n) \\ &\leq 2^{p/2+3d} p^{p/2} (p+2)^{-d} (p+4)^{-d} n^{-p/2}. \end{aligned}$$

If $p < 4d$, then

$$\begin{aligned} \text{av}_p(n, d)^p &\leq 2^{p+d} n^{-p} \sum_{k=1}^{p/2} [(k+1)(k+2)]^{p/4-d} \#(p/2, k, d) \\ &\leq 2^{5p/4} 3^{p/4-d} n^{-p/2}. \end{aligned}$$

Thus we have shown the following theorem:

Theorem 3. *Let p be an even integer. If $p \geq 4d$, then*

$$\text{av}_p(n, d) \leq 2^{1/2+3d/p} p^{1/2} (p+2)^{-d/p} (p+4)^{-d/p} n^{-1/2}.$$

If $p < 4d$, then the estimate $\text{av}_p(n, d) \leq 2^{5/4} 3^{1/4-d} n^{-1/2}$ holds.

For a general $p \in [2, \infty)$ we find a $k \in \mathbb{N}$ with $2k \leq p < 2(k+1)$. Hence there exists a $t \in (0, 1]$ with $1/p = t/2k + (1-t)/2(k+1)$ and from Hölder's inequality we get

$$\text{av}_p(n, d) \leq \text{av}_{2k}(n, d)^t \text{av}_{2(k+1)}(n, d)^{1-t}.$$

Remark 4. The probabilistic argument we used for deriving our upper bound for the average L^p -extreme discrepancy was sketched in [11]. Unfortunately the derivation there contains an error (the number $\sharp(p/2, k, n)$ that appears there has to be substituted by the number $T(p, k, n)$ defined above). For that reason we state here the bounds for the average L^p -star discrepancy $\text{av}_p^*(n, d)$ that we get by mimicking the approach discussed in this section: With the symmetrization argument and (1) we obtain

$$\text{av}_p^*(n, d)^p \leq \left(\frac{2}{n}\right)^p \sum_{k=1}^{p/2} \frac{k^{p/2}}{(k+1)^d} \sharp(p/2, k, n).$$

If $p < 2d$, then $\text{av}_p^*(n, d) \leq 2^{3/2-d/p} n^{-1/2}$. If $p \geq 2d$, then

$$\text{av}_p^*(n, d) \leq 2^{1/2+d/p} p^{1/2} (p+2)^{-d/p} n^{-1/2}. \quad (2)$$

3 Application to the L^∞ -discrepancy

Now we want to derive an upper bound for the inverse $n_\infty(\varepsilon, d)$ of the L^∞ -extreme discrepancy in terms of the average L^p -extreme discrepancy $\text{av}_p(n, d)$. Therefore we define first a “homogeneous version” of the L^∞ -extreme discrepancy: For any $h \in (0, 1]$ and any $t_1, \dots, t_n \in \mathbb{R}^d$ let

$$D_\infty^h(t_1, \dots, t_n) = \inf_{c>0} \sup_{-\mathbf{h} \leq \underline{x} < \bar{x} \leq \mathbf{h}} \left| \prod_{l=1}^d x_l - c \sum_{i=1}^n 1_{[\underline{x}, \bar{x})}(t_i) \right|.$$

Obviously $D_\infty^h(ht_1, \dots, ht_n) = h^d D_\infty^1(t_1, \dots, t_n)$. Further quantities of interest are

$$D_\infty^1(n, d) = \inf_{t_1, \dots, t_n \in [-1, 1]^d} D_\infty^1(t_1, \dots, t_n)$$

and

$$n_\infty^1(\varepsilon, d) := \min\{n \in \mathbb{N} \mid D_\infty^1(n, d) \leq \varepsilon\}.$$

Lemma 5. *For every $\varepsilon > 0$ we have $n_\infty^1(\varepsilon, d) \leq n_\infty(\varepsilon, d) \leq n_\infty^1(\varepsilon/2, d)$.*

The Lemma can be verified by just mimicking the proof of [4, Lemma 2].

Now define for $1 > \varepsilon > 0$, $h = (1 + \varepsilon)^{-1/d}$ and all even natural numbers p

$$A_p^d(\varepsilon) := h^{d(p+2)} \int_{[-\mathbf{1}, (1-2(1-\varepsilon)^{1/d})\mathbf{1}]} \int_{[(1-\varepsilon)^{1/d}\mathbf{1}, \frac{1}{2}(\mathbf{1}-y)]} \left((\varepsilon - 1) + \prod_{j=1}^d z_j \right)^p dz dy$$

and

$$B_p^d(\varepsilon) := \int_{[-\mathbf{1}, -\mathbf{h}]} \int_{[\mathbf{h}, \mathbf{1}]} \left(1 - \prod_{l=1}^d x_l \right)^p 2^{-d} d\bar{x} d\underline{x}.$$

Theorem 6. Let $\varepsilon \in (0, 1)$. If $\varepsilon < D_\infty^1(n, d)$, then we obtain for all even p the inequality $\text{av}_p(n, d) > \min(A_p^d(\varepsilon), B_p^d(\varepsilon))^{1/p}$. Therefore

$$n_\infty^1(\varepsilon, d) \leq \min\{n \mid \exists p \in 2\mathbb{N} : \text{av}_p(n, d) \leq \min(A_p^d(\varepsilon), B_p^d(\varepsilon))^{1/p}\}.$$

Proof. To verify the theorem, we modify the proof from [4, Thm. 6]: Let $D_\infty^1(n, d) > \varepsilon$. For $h \in (0, 1]$ and $t_1, \dots, t_n \in [-1, 1]^d$ we have

$$D_\infty^h(t_1, \dots, t_n) = h^d D_\infty^1(t_1/h, \dots, t_n/h) > \varepsilon h^d.$$

Therefore we find $\underline{x}, \bar{x} \in [-h, h]^d$ with $\underline{x} < \bar{x}$ and

$$\left| \prod_{l=1}^d x_l - \frac{1}{n} \sum_{i=1}^n 1_{[\underline{x}, \bar{x}]}(t_i) \right| > \varepsilon h^d.$$

Case 1: There holds

$$\prod_{l=1}^d x_l - \frac{1}{n} \sum_{i=1}^n 1_{[\underline{x}, \bar{x}]}(t_i) > \varepsilon h^d.$$

With respect to its volume the box $[\underline{x}, \bar{x}]$ contains not sufficiently many sample points. This holds also for slightly smaller boxes. If $[\underline{v}, \bar{v}] \subseteq [\underline{x}, \bar{x}]$, then

$$\prod_{j=1}^d v_j - \frac{1}{n} \sum_{i=1}^n 1_{[\underline{v}, \bar{v}]}(t_i) > \varepsilon h^d - \prod_{j=1}^d x_j + \prod_{j=1}^d v_j.$$

This leads to

$$\begin{aligned} D_p(t_1, \dots, t_n)^p &> \int_{[\underline{x}, \bar{x}]} \int_{[\underline{v}, \bar{v}]} \left(\varepsilon h^d - \prod_{j=1}^d x_j + \prod_{j=1}^d v_j \right)_+^p 2^{-d} d\bar{v} dv \\ &= \int_{[-\mathbf{h}, -\mathbf{h}+2x]} \int_{[\underline{z}+2(\mathbf{h}-x), \mathbf{h}]} \left(\varepsilon h^d - \prod_{j=1}^d x_j + \prod_{j=1}^d (z_j + x_j - h) \right)_+^p 2^{-d} d\bar{z} d\underline{z}, \end{aligned}$$

where in the last step we made a change of coordinates: $\underline{z} = \underline{v} - \underline{x} - \mathbf{h}$ and $\bar{z} = \bar{v} - \bar{x} + \mathbf{h}$. If we translate edge points v and w , $v \leq w$, of anchored boxes $[0, v)$ and $[0, w)$ by a vector $a \geq 0$, then it is a simple geometrical observation that the volumes of the corresponding anchored boxes satisfy

$$\text{vol}([0, w)) - \text{vol}([0, v)) \leq \text{vol}([0, w + a)) - \text{vol}([0, v + a)).$$

In particular, if $w = x$, $v = z + x - \mathbf{h}$ and $a = \mathbf{h} - x$, then

$$\prod_{j=1}^d x_j - \prod_{j=1}^d (z_j + x_j - h) \leq h^d - \prod_{j=1}^d z_j.$$

This, and integrating over the variable z instead over $\bar{z}$, leads to

$$D_p(t_1, \dots, t_n)^p > \int_{[-\mathbf{h}, -\mathbf{h}+2\mathbf{x}]} \int_{[\mathbf{h}-\mathbf{x}, \frac{1}{2}(\mathbf{h}-\underline{z})]} \left((\varepsilon - 1)h^d + \prod_{j=1}^d z_j \right)_+^p dz d\underline{z}.$$

We can ignore those vectors z with a component $z_i < (1 - \varepsilon)h$, since they satisfy the relation $(\varepsilon - 1)h^d + \prod_{j=1}^d z_j < 0$. As $x_i > \varepsilon h$ for all $1 \leq i \leq d$, we get

$$\begin{aligned} D_p(t_1, \dots, t_n)^p &> \int_{[-\mathbf{h}, (2\varepsilon-1)\mathbf{h}]} \int_{[(1-\varepsilon)\mathbf{h}, \frac{1}{2}(\mathbf{h}-\underline{z})]} \left((\varepsilon - 1)h^d + \prod_{j=1}^d z_j \right)_+^p dz d\underline{z} \\ &\geq \int_{[-\mathbf{h}, (1-2(1-\varepsilon)^{1/d})\mathbf{h}]} \int_{[(1-\varepsilon)^{1/d}\mathbf{h}, \frac{1}{2}(\mathbf{h}-\underline{z})]} \left((\varepsilon - 1)h^d + \prod_{j=1}^d z_j \right)_+^p dz d\underline{z}. \end{aligned}$$

Case 2: There holds

$$\frac{1}{n} \sum_{i=1}^n 1_{[\underline{x}, \bar{x})}(t_i) - \prod_{l=1}^d x_l > \varepsilon h^d.$$

The box $[\underline{x}, \bar{x})$ contains too many points, and this is also true for somewhat larger boxes. If $[\underline{x}, \bar{x}) \subseteq [\underline{w}, \bar{w})$, then

$$\frac{1}{n} \sum_{i=1}^n 1_{[\underline{w}, \bar{w})}(t_i) - \prod_{l=1}^d w_l > \varepsilon h^d + \prod_{l=1}^d x_l - \prod_{l=1}^d w_l.$$

This implies

$$\begin{aligned} D_p(t_1, \dots, t_n)^p &> \int_{[-1, \underline{x}]} \int_{[\bar{x}, 1]} \left(\varepsilon h^d + \prod_{l=1}^d x_l - \prod_{l=1}^d w_l \right)_+^p 2^{-d} d\bar{w} d\underline{w} \\ &\geq \int_{[-1-\mathbf{h}-\underline{x}, -\mathbf{h}]} \int_{[\mathbf{h}, 1+\mathbf{h}-\bar{x}]} \left(\varepsilon h^d + \prod_{l=1}^d x_l - \prod_{l=1}^d (z_l - h + x_l) \right)_+^p 2^{-d} d\bar{z} d\underline{z}, \end{aligned}$$

where we made the substitutions $\bar{z} = \bar{w} - \bar{x} + \mathbf{h}$ and $\underline{z} = \underline{w} - \underline{x} - \mathbf{h}$. If we restrict the domain of integration and use the simple geometric observation mentioned in the discussion of Case 1, we obtain

$$D_p(t_1, \dots, t_n)^p > \int_{[-1, -\mathbf{h}]} \int_{[\mathbf{h}, 1]} \left((1 + \varepsilon)h^d - \prod_{l=1}^d z_l \right)_+^p 2^{-d} d\bar{z} d\underline{z}.$$

If we choose $h = (1 + \varepsilon)^{-1/d}$, then $D_p(t_1, \dots, t_n)^p > B_p^d(\varepsilon)$.

Our analysis results in $D_p(t_1, \dots, t_n)^p > \min\{A_p^d(\varepsilon), B_p^d(\varepsilon)\}$ for all $t_1, \dots, t_n \in [-1, 1]^d$. Theorem 6 follows now by integration. $\square$

Lemma 7. Let $\varepsilon \in (0, 1/2]$ and $p \geq 4d$ be an even integer. Then

$$\min(A_p^d(\varepsilon), B_p^d(\varepsilon))^{1/p} \geq \frac{1}{3}\varepsilon\left(\frac{\varepsilon}{4d}\right)^{2d/p}.$$

Proof. Let again $h = (1 + \varepsilon)^{-1/d}$. From the definition of $B_p^d(\varepsilon)$ follows

$$\begin{aligned} B_p^d(\varepsilon) &\geq \int_{[-(1+\varepsilon/2)^{-1/d}\mathbf{1}, -\mathbf{h}]} \int_{[\mathbf{h}, (1+\varepsilon/2)^{-1/d}\mathbf{1}]} (1 - (1 + \varepsilon/2)^{-1})^p 2^{-d} d\bar{x} d\underline{x} \\ &= 2^{-d}(1 - (1 + \varepsilon/2)^{-1})^p ((1 + \varepsilon/2)^{-1/d} - (1 + \varepsilon)^{-1/d})^{2d}. \end{aligned}$$

As $\varepsilon \leq 1/2$, it is straightforward to verify the inequalities $1 - (1 + \varepsilon/2)^{-1} \geq 2\varepsilon/5$ and $(1 + \varepsilon/2)^{-1/d} - (1 + \varepsilon)^{-1/d} \geq \varepsilon/4d$. That implies

$$B_p^d(\varepsilon)^{1/p} \geq 2^{-d/p} \frac{2}{5} \varepsilon \left(\frac{\varepsilon}{4d}\right)^{2d/p} \geq 2^{-1/4} \frac{2}{5} \varepsilon \left(\frac{\varepsilon}{4d}\right)^{2d/p} \geq \frac{1}{3} \varepsilon \left(\frac{\varepsilon}{4d}\right)^{2d/p}.$$

We can estimate $A_p^d(\varepsilon)$ in the following way:

$$\begin{aligned} A_p^d(\varepsilon) &\geq h^{d(p+2)} \int_{[-\mathbf{1}, (1-2(1-\varepsilon/2)^{1/d})\mathbf{1}]} \int_{[(1-\varepsilon/2)^{1/d}\mathbf{1}, \frac{1}{2}(1-y)]} (\varepsilon/2)^p dz dy \\ &= (1 + \varepsilon)^{-p-2} (\varepsilon/2)^p (1 - (1 - \varepsilon/2)^{1/d})^{2d}. \end{aligned}$$

Since $1 - (1 - \varepsilon/2)^{1/d} \geq \varepsilon/2d$, we get

$$\begin{aligned} A_p^d(\varepsilon)^{1/p} &\geq \frac{1}{(1 + \varepsilon)^{1+2/p}} \frac{\varepsilon}{2} \left(\frac{\varepsilon}{2d}\right)^{2d/p} \\ &\geq \frac{1}{1 + \varepsilon} \left(\frac{2^d}{1 + \varepsilon}\right)^{2/p} \frac{\varepsilon}{2} \left(\frac{\varepsilon}{4d}\right)^{2d/p} \geq \frac{1}{3} \varepsilon \left(\frac{\varepsilon}{4d}\right)^{2d/p}. \end{aligned}$$

□

Let now $k \in \mathbb{N}$, $p = 4kd$ and $\varepsilon \in (0, 1/2)$. With Theorem 3 and Lemma 7 it is easily verified that

$$n \geq 9 \cdot 2^{3(1+1/2k)} k^{1-1/k} d\varepsilon^{-2-1/k} \quad \text{ensures} \quad \text{av}_p(n, d) \leq \min(A_p^d(\varepsilon), B_p^d(\varepsilon))^{1/p}.$$

This, Lemma 5 and Theorem 6 lead to the following theorem:

Theorem 8. Let $\varepsilon \in (0, 1/2)$ and $k \in \mathbb{N}$. Then $n_\infty(\varepsilon, d) \leq C_k d\varepsilon^{-2-1/k}$, where the constant C_k is bounded from above by $9 \cdot 2^{5(1+1/2k)} k^{1-1/k}$.

Remark 9. In a similar way we can use the bound for the average L^p -star discrepancy to calculate an upper bound for the inverse $n_\infty^*(d, \varepsilon)$ of the star discrepancy: With (2), [4, Thm. 6] and [4, Lemma 3] (where we can replace the factor $\sqrt{2/3}$ by 1—cf. with the proof of Lemma 7), we obtain

$$n_\infty^*(d, \varepsilon) \leq 9 \cdot 2^{4+3/k} k^{1-1/k} d\varepsilon^{-2-1/k}. \quad (3)$$

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