

A remark on the minimal dispersion

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Such an ε is called dispersion (of the cube) or *minimal dispersion* and denoted by

$$\text{disp}^*(n, d).$$

Notations.

Consider the set of all axis parallel boxes contained in the cube $[0, 1]^d$,

$$\mathcal{R}_d := \left\{ \prod_{i=1}^d I_i \mid I_i = [a_i, b_i) \subset [0, 1] \right\}.$$

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In this talk it will be more convenient to work with its inverse, the function

$$N(\varepsilon, d) = \min\{n \in \mathbb{N} \mid \text{disp}^*(n, d) \leq \varepsilon\}.$$

Known results.

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The first bound showing that the minimal dispersion grows with the dimension was obtained by Aistleitner–Hinrichs–Rudolf (17):

$$N(\varepsilon, d) \geq (1 - 4\varepsilon) \frac{\log_2 d}{4\varepsilon}.$$

Known results.

$$(1 - 4\varepsilon) \frac{\log_2 d}{4\varepsilon} \leq N(\varepsilon, d) \leq \frac{Cd^2 \log d}{\varepsilon}. \quad (1)$$

Thus, when the dimension d is fixed and $\varepsilon \rightarrow 0$ the problem is essentially solved:

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Blumer–Ehrenfeucht–Haussler–Warmuth (89) provided a general bound in terms of VC dimension of \mathcal{R}_d . Using that this dimension equals to $2d$,

$$N(\varepsilon, d) \leq \frac{Cd}{\varepsilon} \log_2 \left(\frac{C}{\varepsilon} \right). \quad (2)$$

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Estimate (2) is better than the upper bound in (1) in the regime $\varepsilon \geq d^{-Cd}$.

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A natural conjecture: $N(\varepsilon, d) \approx \frac{d}{\varepsilon}$. (Bukh–Chao: $N(\varepsilon, d) \geq \frac{d}{e\varepsilon}$ if $\varepsilon \leq (4d)^{-d}$.)

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Their upper bound is better in the regime

$$\varepsilon \geq \frac{C (\log_2 d)^2}{d}.$$

Known results.

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If $\varepsilon \in [1/2, 1]$ then $N(\varepsilon, d) = 1$, indeed one can take the point $(1/2, 1/2, \dots, 1/2)$.

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Sosnovec (18):
$$N(\varepsilon, d) \leq 1 + \left\lfloor \frac{1}{\varepsilon - 1/4} \right\rfloor \quad \text{for } \varepsilon > 1/4.$$

Does not grow when $d \rightarrow \infty$. Recall, for $\varepsilon < 1/4$, we have $N(\varepsilon, d) \geq C_\varepsilon \log_2 d$.

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Problem. What is $N(1/4, d)$? How does $N(\varepsilon, d)$ behave when $\varepsilon \rightarrow (1/4)^\pm$.

Upper bounds: summary

Till very recently:

$$N(\varepsilon, d) \leq \begin{cases} \frac{C \ln d}{\varepsilon^2} \ln^2 \left(\frac{1}{\varepsilon} \right), & \text{if } \varepsilon \geq \frac{\ln^2 d}{d}, \\ \frac{C d}{\varepsilon} \ln \left(\frac{1}{\varepsilon} \right), & \text{if } \frac{\ln^2 d}{d} \geq \varepsilon \geq \exp(-C^d), \\ \frac{C^d}{\varepsilon}, & \text{if } \varepsilon \leq \exp(-C^d). \end{cases}$$

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Using the [Bukh–Chao](#) result:

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Theorem (small ε)

Let $d \geq 2$ and $\varepsilon \leq 1/2$. Then

$$(i) \quad N(\varepsilon, d) \leq \frac{C \ln d}{\varepsilon} \ln \left(\frac{1}{\varepsilon} \right), \quad \text{provided that } \varepsilon \leq \exp(-d),$$

$$(ii) \quad N(\varepsilon, d) \leq \frac{C d}{\varepsilon} \ln \ln \left(\frac{2}{\varepsilon} \right), \quad \text{provided that } \varepsilon \geq \exp(-d).$$

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Moreover, the random choice of points uniformly distributed on $[0, 1]^d$ works.

Thus, for $\varepsilon \leq \exp(-d)$ we have

$$\frac{\ln d}{6\varepsilon} \leq N(\varepsilon, d) \leq \frac{C \ln d}{\varepsilon} \ln \left(\frac{1}{\varepsilon} \right),$$

In (ii) the improvement is only in substitution of $\ln(1/\varepsilon)$ with $\ln \ln(1/\varepsilon)$.

New bounds.

Hinrichs–Krieg–Kunsch–Rudolf (20): for a random choice of points

$$\max \left\{ \frac{c}{\varepsilon} \ln \left(\frac{1}{\varepsilon} \right), \frac{d}{2\varepsilon} \right\} \leq N_{ran}(\varepsilon, d)$$

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Theorem (relatively large ε)

Let $d \geq 2$ and $\frac{\ln d}{d} \leq \varepsilon \leq 1/2$. Then

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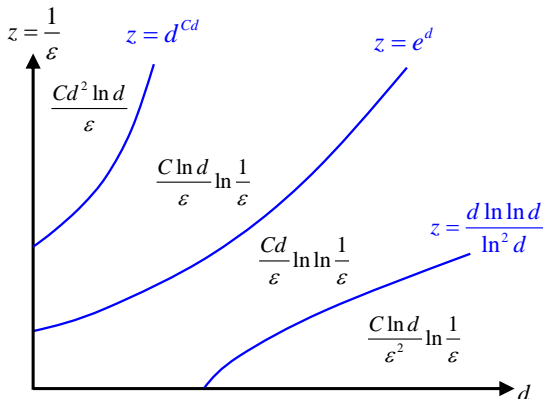
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The prove also uses random points, but one needs to adjust the distribution.

State of the art.

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Some ideas of the proof.

Consider a set P of random points independently and uniformly distributed in the cube. To show that (with good probability) every box of volume ε contains at least one point from P , we need to construct a set \mathcal{N} of “test boxes.” It should satisfy

each rectangle in \mathcal{N} contains a point from P \implies

each rectangle in \mathcal{R}_d of volume at least ε contains a point from P .

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We use a different approach, which fits better this problem.

Some ideas of the proof.

Denote

$$\mathcal{B}_{\varepsilon,d} := \left\{ B \in \mathcal{R}_d \mid |B| \geq \varepsilon \right\}.$$

Definition. We say that $\mathcal{N} \subset \mathcal{R}_d$ is a δ -net for $\mathcal{B}_{\varepsilon,d}$ if

$$\forall B \in \mathcal{B}_{\varepsilon,d} \exists B_0 \in \mathcal{N} : \quad B_0 \subset B \quad \text{and} \quad |B_0| \geq (1 - \delta)|B|.$$

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Denote

$$\mathcal{B}_{\varepsilon,d} := \left\{ B \in \mathcal{R}_d \mid |B| \geq \varepsilon \right\}.$$

Definition. We say that $\mathcal{N} \subset \mathcal{R}_d$ is a δ -net for $\mathcal{B}_{\varepsilon,d}$ if

$$\forall B \in \mathcal{B}_{\varepsilon,d} \exists B_0 \in \mathcal{N} : \quad B_0 \subset B \quad \text{and} \quad |B_0| \geq (1 - \delta)|B|.$$

A variant of the following lemma using random points and the union bound was proved by [Rudolf](#).

Lemma (size of a net)

Let $d \geq 1$ and $\varepsilon, \delta \in (0, 1)$. Let \mathcal{N} be a δ -net for $\mathcal{B}_{\varepsilon,d}$ with $|\mathcal{N}| \geq 3$. Then with probability at least $1 - 1/|\mathcal{N}|$

$$N = \left\lceil \frac{3 \ln |\mathcal{N}|}{(1 - \delta)\varepsilon} \right\rceil.$$

a random choice of N points satisfies the desired property.

Proof of the lemma.

Consider N independent random points X_1, \dots, X_N uniformly drawn from $[0, 1]^d$.
It is enough to show that

$$\forall B \in \mathcal{N} \text{ with } |B| \geq v = (1 - \delta)\varepsilon \quad \exists j \leq N : X_j \in B.$$

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Fix such a box B . Using independence of X_j 's,

$$\mathbb{P}(\{\forall j \leq N : X_j \notin B\}) = (1 - v)^N < \exp(-vN).$$

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Thus, as far as

$$|\mathcal{N}| \exp(-vN) \leq 1$$

there exists a realization of X_j 's with the desired property.

Construction of a net.

Dispersion on the torus.

We consider periodic axis parallel boxes, that is, boxes of the form

$$\prod_{i=1}^d I_i(a_i, b_i), \quad a_i, b_i \in [0, 1],$$

where

$$I_i(a, b) := \begin{cases} (a_i, b_i), & \text{whenever } 0 \leq a_i < b_i \leq 1, \\ [0, 1] \setminus [b_i, a_i], & \text{whenever } 0 \leq b_i < a_i \leq 1. \end{cases}$$

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Combining recent results of [M. Ullrich \(18\)](#) and [Rudolf \(18\)](#)

$$\frac{d}{\varepsilon} \leq \tilde{N}(\varepsilon, d) \leq \frac{8d}{\varepsilon} \left(\ln d + \ln \left(\frac{8}{\varepsilon} \right) \right).$$

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Note that the VC dimension of the set of periodic axis parallel boxes is of the order $d \ln d$, therefore the [Blumer–Ehrenfeucht–Haussler–Warmuth](#) result leads to

$$\tilde{N}(\varepsilon, d) \leq \frac{8d \ln d}{\varepsilon} \ln \left(\frac{8}{\varepsilon} \right)$$

— worse than the [Rudolf](#) bound.

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We improve the **Rudolf** upper bound in the case $\varepsilon \leq 1/d^C$.

Theorem (bounds in the periodic case)

Let $d \geq 2$ and $\varepsilon \in (0, 1/2]$. Then

$$(i) \quad \tilde{N}(\varepsilon, d) \leq \frac{C \ln d}{\varepsilon} \ln \left(\frac{1}{\varepsilon} \right), \quad \text{provided that } \varepsilon \leq \exp(-d),$$

$$(ii) \quad \tilde{N}(\varepsilon, d) \leq \frac{C d \ln d}{\varepsilon}, \quad \text{provided that } \varepsilon \geq \exp(-d).$$

Moreover, the random choice of points uniformly distributed on $[0, 1]^d$ works.