Measures of $\epsilon$-complexity

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Abstract

We study some measures which are related to the notion of the $\epsilon$-complexity. We prove that measure of $\epsilon$-complexity defined on the base of the notion of $\epsilon$-separability is equivalent to the dual measure that is defined through $\epsilon$-nets.

Keywords: Complexity, Separability, Bernoulli measure

1 Introduction

The problems under consideration in this article were originated in the process of study of complexity of behavior of orbits in dynamical systems. While symbolic complexity (see, for instance [4]) deals with symbolic systems and topological complexity ([2]) reflects pure topological features of dynamics, the $\epsilon$-complexity depends essentially on a distance in the phase space (see definition below). If one has a dynamical system generated by a continuous map $f : X \to X$ where $X$ is a metric space with a distance $\rho$, one can introduce the sequence of distances ([3])

$$\rho_n(x, y) = \max_{0 \leq i \leq n-1} \rho(f^i x, f^i y), \quad n \in \mathbb{N},$$

and study the $\epsilon$-complexity with respect to the distance $\rho_n$ as a function of “time” $n$. This function reflects the evolution of instability of orbits in time [1]. But to study it in details, one needs to know more about general properties of the $\epsilon$-complexity of a metric space (without dynamics).

The goal this article is to introduce and study quantities which contain an essential information about $\epsilon$-complexity, the measures of $\epsilon$-complexity in an “abstract” metric space. The main results will be related to the $\epsilon$-complexity defined on the base of the notion of $\epsilon$-separability. The notion was used first by Kolmogorov and Tikhomirov [9] in their study of solutions of PDE and realization of random processes (Shannon suggested to pay attention to this notions in 1949, though). We will also study $\epsilon$-complexities based on the notion of $\epsilon$-nets. We prove that measure of $\epsilon$-complexity defined on the base of the notion of $\epsilon$-separability is equivalent to the dual measure that is defined through $\epsilon$-nets.

It appeared naturally that some results and ideas from discrete mathematics are worth to be exploited. We believe that we made the first step in this direction.

2 Set-up and definitions

2.1 Separated sets and complexity

Let $X$, $d$ be a compact metric space with a distance $d$.

Definition 1 1. Given $\epsilon > 0$, a set $Y \subseteq X$ is $\epsilon$-separated iff for any different $x, y \in Y$ one has $d(x, y) \geq \epsilon$.

2. The number

$$C_\epsilon(X, d) = C_\epsilon := \max\{|Y|, \ Y \text{ is an } \epsilon \text{-separated set}\},$$

where $|\cdot|$ denotes the cardinality of a set, is called the $\epsilon$-complexity of $X$.

3. An $\epsilon$-separated set $Y$ is optimal iff $|Y| = C_\epsilon$.

Let us show the following natural inequality.
Proposition 1 Given $D_1, D_2 \subseteq X$ and $\epsilon > 0$ one has
\[ C_\epsilon(D_1 \cup D_2) \leq C_\epsilon(D_1) + C_\epsilon(D_2). \]

Proof. Let $Y \subseteq D_1 \cup D_2$ be an optimal $\epsilon$-separated set in $D_1 \cup D_2$. Then $Y = Y \cap D_i$ is an $\epsilon$-separated set in $D_i$ and $|Y| \leq |Y_1| + |Y_2| \leq C_\epsilon(D_1) + C_\epsilon(D_2)$. □

Remark. Invariant sets in dynamical systems can be treated as results of inductive procedures. For example, the dynamical system generated by the map $f : \mathbb{R} \rightarrow \mathbb{R}$,
\[ f(x) = \begin{cases} 3x, & x \leq 1/2, \\ 3x - 3, & x > 1/2, \end{cases} \]
has an invariant set $K$ containing all orbits belonging to the interval $[0, 1]$. One can see that $K$ is the one-third Cantor set, so that
\[ K = \bigcap_{n=1}^{\infty} \bigcup_{(i_0, \ldots, i_n-1)} \Delta_{i_0\ldots i_n-1}, \]
where $i_j \in \{0, 1\}$. $\Delta_{i_0\ldots i_n-1}$ are intervals of the length $3^{-n}$ arising on the $n$-th step of construction of the Cantor set. Therefore, if $\epsilon \approx 3^{-n}$ then $C_\epsilon \approx 2^n = \{\text{the number of different words of length } n \text{ in the full shift with 2 symbols}\} = e^h_n$, where $h = \ln 2$ is the topological entropy of the full shift. Thus,
\[ \frac{\ln C_\epsilon}{\ln \epsilon} \approx \frac{\ln 2}{\ln 1/3} = \frac{h}{\ln \lambda} = \dim_H K, \]
where $\dim_H K$ is the Hausdorff dimension of $K$ and $\lambda = 1/3$ is the contraction coefficient. We obtained the familiar Furstenberg formula.

This example shows that if a subset of a metric space is the result of an inductive procedure governed by a symbolic dynamical system then the $\epsilon$-complexity contains, in fact, an important dynamical information.

2.2 $\epsilon$-nets and complexity

In this subsection we give a dual definition of complexity. Given $x \in X$ let $O_\epsilon(x) = \{y : d(x, y) < \epsilon\}$, the ball of radius $\epsilon$ centered at $x$. Given $Y \subseteq X$ let $O_\epsilon(Y) = \bigcup_{x \in Y} O_\epsilon(x)$.

Definition 2 1. Given $\epsilon > 0$, a set $Y \subseteq X$ is an $\epsilon$-net iff $O_\epsilon(Y) = X$.

2. The number
\[ R_\epsilon(X, d) = R_\epsilon := \min\{|Y| : Y \text{ is an } \epsilon\text{-net}\}, \]
is called the dual $\epsilon$-complexity of $X$.

3. An $\epsilon$-net $Y$ is optimal iff $|Y| = R_\epsilon$.

The similar results to the one in Proposition holds for dual complexities.

Proposition 2 Given $D_1, D_2 \subseteq X$ and $\epsilon > 0$ one has
\[ R_\epsilon(D_1 \cup D_2) \leq R_\epsilon(D_1) + R_\epsilon(D_2). \]

Proof. Let $Y_i \subseteq D_i$ be an optimal $\epsilon$-net in $D_i$. Then $Y = Y_1 \cup Y_2$ is an $\epsilon$-net in $D_1 \cup D_2$ and $R_\epsilon(D_1 \cup D_2) \leq |Y| \leq |Y_1| + |Y_2| = R_\epsilon(D_1) + R_\epsilon(D_2)$. □

Any optimal $\epsilon$-separated set is an $\epsilon$-net, therefore $C_\epsilon \geq R_\epsilon$. On the other hand the following statement holds.

Proposition 3 $R_{\epsilon/2} \geq C_\epsilon$

Proof. It follows directly from the definition that any pair of different points in an $\epsilon$-separated set $Z$ cannot belong to a ball of radius $\epsilon/2$. Thus we cannot cover $Z$ by less than $|Z|$ balls of radius $\epsilon/2$. Assuming that $Z$ is optimal we obtain the inequality above. □

Let us introduce
\[ b_\epsilon = \sup_{x \in X} R_{\epsilon/2}(O_\epsilon(x)). \]
Obviously, for any $D \subseteq X$ one has $b_\epsilon R_\epsilon(D) \geq R_{\epsilon/2}(D)$. It is not difficult to check that $b_\epsilon \leq 2^\epsilon(2^d + 1)$ for a subset of the Euclidean space $\mathbb{R}^d$. 

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2.3 Ultrafilters

Now we give some known results and definitions that can be found, for instance, in [5].

Definition 3 A set $\mathcal{F} \subseteq 2^\mathbb{N}$ is called to be a filter over $\mathbb{N}$ iff it satisfies the following conditions:

- If $A \in \mathcal{F}$ and $B \in \mathcal{F}$, then $A \cap B \in \mathcal{F}$,
- If $A \in \mathcal{F}$ and $A \subseteq B$ then $B \in \mathcal{F}$,
- $\emptyset \notin \mathcal{F}$.

Let $a_n$ be a sequences of real numbers, $a$ is called to be a limit of $a_n$ with respect to a filter $\mathcal{F}$, $a = \lim_{\mathcal{F}} a_n$, if for any $\varepsilon > 0$ one has \{n | |a_n - a| < \varepsilon\} $\in \mathcal{F}$. From the definition of a filter it follows that $\lim_{\mathcal{F}} a_n$ is unique, if exists.

Example Let $\mathcal{F}_F = \{A \subseteq \mathbb{N} | \mathbb{N}\setminus A$ is finite \}. $\mathcal{F}_F$ is said to be a Frechét filter. One can check that it is, indeed, a filter. A limit with respect to $\mathcal{F}_F$ coincides with ordinary limit.

Definition 4 A filter $\mathcal{F}$ is called to be ultrafilter iff for any set $A \subseteq \mathbb{N}$ one has $A \in \mathcal{F}$ or $\mathbb{N}\setminus A \in \mathcal{F}$.

Theorem 1 A bounded sequences has a limit with respect to an ultrafilter. This limit is unique.

Example For $i \in \mathbb{N}$ let $\mathcal{F}_i = \{A \subseteq \mathbb{N} | i \in A\}$. It is an ultrafilter. Such an ultrafilter is called proper for $i$. One can check that $\lim_{\mathcal{F}_i} a_n = a_i$. So, limits with respect to a proper ultrafilter are not interesting.

Proposition 4 An ultrafilter $\mathcal{F}$ is proper (for some $i \in \mathbb{N}$) if and only if it contains a finite set.

This proposition implies that an ultrafilter is non-proper if and only if it is an extension of the Frechét filter $\mathcal{F}_F$. On the other hand, it follows from the Zorn lemma that any filter can be extended to an ultrafilter.

Proposition 5 There is an ultrafilter $\mathcal{F} \supset \mathcal{F}_F$. Any such an ultrafilter is non-proper.

3 Measures of complexity

Our goal is to define a measure reflecting an asymptotic behavior of the $\varepsilon$-complexity as $\varepsilon$ goes to 0. For that we will use the technique of ultrafilters.

Given $\varepsilon > 0$, consider an optimal $\varepsilon$-separated set $A_{\varepsilon}$. Introduce the following functional

$$I_\varepsilon(\phi) = \frac{1}{C_\varepsilon} \sum_{x \in A_{\varepsilon}} \phi(x)$$

where $\phi : X \to R$ is a continuous function. It is clear that $I_\varepsilon$ is a positive bounded linear functional on $C(X)$. Moreover, for any $\phi \in C(X)$ the family $I_\varepsilon(\phi)$ is bounded. Fix a sequence $E = \{\varepsilon_n\}$, $\varepsilon_n \to 0$ as $n \to \infty$ and an arbitrary non-proper ultrafilter $\mathcal{F}$. Consider

$$I(\phi) = \lim_{\mathcal{F}} I_{\varepsilon_n}(\phi).$$

$I$ is a positive bounded linear functional on $C(X)$.

Theorem 2 The functional $I$ is independent of the choice of an optimal sets $A_{\varepsilon}$.

Proof. The proof is based on the following proposition.

Proposition 6 Let $A$ and $B$ be optimal $\varepsilon$-separated sets. There exists a one-to-one map $\alpha : A \to B$ such that $d(x, \alpha(x)) \leq \varepsilon$ for any $x \in A$.

Let $A_{\varepsilon}$ and $B_{\varepsilon}$ be optimal $\varepsilon$-separated sets, $\varepsilon \in E$. Let $\alpha_{\varepsilon} : A_{\varepsilon} \to B_{\varepsilon}$ be the map from Proposition 6. Then

$$\left| \frac{1}{C_\varepsilon} \sum_{x \in A_{\varepsilon}} \phi(x) - \frac{1}{C_\varepsilon} \sum_{x \in B_{\varepsilon}} \phi(x) \right| = \frac{1}{C_\varepsilon} \sum_{x \in A_{\varepsilon}} |(\phi(x) - \phi(\alpha_{\varepsilon}(x)))| \leq r_\varepsilon(\varepsilon)$$

where $r_\varepsilon(\varepsilon) = \sup \{|\phi(x) - \phi(y)| : d(x, y) < \varepsilon\}$, the modulus of continuity of $\phi$. Since $X$ is a compact, $r_\varepsilon(\varepsilon) \to 0$ as $\varepsilon \to 0$. It implies the desired result due to the choice of the ultrafilter $\mathcal{F}$. So, we need only to prove Proposition 6 it will be done below. □

In the proof of Proposition 6 we will need the Marriage Lemma of P. Hall, see for instance [10].

Lemma 1 For an indexed collections of finite sets $F_1, F_2, \ldots, F_k$ the following conditions are equivalent:

- there exists an injective function $\alpha : \{1, 2, \ldots, k\} \to \bigcup_{i=1}^k F_i$ such that $\alpha(i) \in F_i$;
For all $S \subseteq \{1, 2, \ldots, k\}$ one has $| \bigcup_{i \in S} F_i | \geq |S|$. 

Recall that $O_\epsilon(x) = \{ y : d(x, y) < \epsilon \}$, the ball of radius $\epsilon$ centered at $x$. Given $Y \subseteq X$ let $O_\epsilon(Y) = \bigcup_{x \in Y} O_\epsilon(x)$.

**Proof of Proposition 8**

For any $x \in A$ let $B_x = O_\epsilon(x) \cap B$. If we show that for any $S \subseteq A$ the following inequality holds

$$| \bigcup_{x \in S} B_x | \geq |S|,$$

then the proposition follows from Lemma 1 due to $|A| = |B| = C\epsilon$. To prove inequalities (1), suppose that $| \bigcup_{x \in S} B_x | = |O_\epsilon(S) \cap B| < |S|$ for some $S \subseteq A$. Then

$$(S \cup (B \setminus (O_\epsilon(S) \cap B)) = |S| + (|B| - |O_\epsilon(S) \cap B|) > |B| = C\epsilon,$$

on the other hand, the set $S \cup (B \setminus (O_\epsilon(S) \cap B))$ is $\epsilon$-separated. We have a contradiction with optimality of $B$. □

So, we have defined a functional $I$ which may depend on the choice of the sequence $E$ and the ultrafilter $F$ only. Sometimes we will write $I_{E,F}$ to emphasize this dependence. It is well known, that $I_{E,F}$ generate unique regular Borel measure $\mu_{E,F}$ on $X$ such that $\mu_{E,F}(X) = 1$.

**Definition 5** The measures $\mu_{E,F}(X)$ will be called measures of complexity.

We are going to show examples of $(X, d)$ when $\mu_{E,F} = \mu$ is independent on $E, F$ and when $\mu_{E,F}$ depends on $E, F$. In the first case

$$I_{E,F}(\phi) = I(\phi) = \lim_{\epsilon \to 0} I_{\epsilon}(\phi).$$

Of course, it is difficult to find optimal sets and construct directly measures of complexity in real situations. Nevertheless, it is possible to work with them by using some of their intrinsic properties. Let us show now that measures of complexity are invariant with respect to local isometries.

**Definition 6** A homeomorphism $\tau : X \to X$ is called to be $\epsilon$-isometry iff $d(x, y) = d(\tau(x), \tau(y))$ for all $x, y \in X, d(x, y) \leq \epsilon$. A homeomorphism $\tau : X \to X$ is called to be local isometry if it is $\epsilon$-isometry for some $\epsilon > 0$.

It is clear that an isometry is a local isometry.

**Proposition 7** Local isometries with composition form a group.

**Proof.** It is easy to check that the composition of two $\epsilon$-isometries is an $\epsilon$-isometry. Let $\tau$ be an $\epsilon$-isometry. Then $\tau^{-1}$ is uniformly continuous and there exists $\epsilon' > 0$ such that if $d(x, y) \leq \epsilon$, then $d(\tau^{-1}(x), \tau^{-1}(y)) \leq \epsilon$. Consequently, if $d(x, y) \leq \epsilon'$ then $d(x, y) = d(\tau^{-1}(x), \tau^{-1}(y))$, so, $\tau^{-1}$ is an $\epsilon'$-isometry. □

We do not know if $\epsilon$-isometries form a group.

**Proposition 8** Let $\tau$ be an $\epsilon_0$-isometry and $A$ be an $\epsilon$-separated set, $\epsilon \leq \epsilon_0$. Then $\tau^{-1}(A)$ is also $\epsilon$-separated.

**Proof.** Assume, on the contrary, that $\tau^{-1}(A)$ is not $\epsilon$-separated, i.e., there are different $x, y \in \tau^{-1}(A)$ with $d(x, y) < \epsilon \leq \epsilon_0$. Then $d(\tau(x), \tau(y)) = d(x, y) < \epsilon$, so $A$ cannot be $\epsilon$-separated. □

**Theorem 3** Let $\tau$ be a local isometry. Then $\mu_{E,F}$ is invariant, i.e. $\mu_{E,F}(A) = \mu_{E,F}(\tau^{-1}(A))$ for all measurable $A$.

**Proof.** It is enough to show that for all $\phi \in C(X)$

$$I_{E,F}(\phi \circ \tau) = I_F(\phi),$$

(2)

There exists $\epsilon_0 > 0$, such that $\tau$ is an $\epsilon_0$-isometry. Let $A_i$ be an optimal $\epsilon$-separated set, $\epsilon \leq \epsilon_0$. It follows from Proposition 8 that $\tau^{-1}(A_i)$ is an optimal $\epsilon$-separated set. It implies the validity of Equation (2).

Indeed,

$$I_{\epsilon}(\phi) = \frac{1}{C_{\epsilon}} \sum_{x \in A_i} \phi(x), \quad I_{\epsilon}(\phi \circ \tau) = \frac{1}{C_{\epsilon}} \sum_{x \in \tau^{-1}(A_i)} \phi(x),$$

and the result follows from Theorem 2. □

**Corollary 1** Let a continuous group operation $\ast$ be defined on $X$ such that right shifts $r_g(x) = g \ast x$ (left shifts $l_g(x) = x \ast g$) are local isometries for all $g \in X$. Then $\mu_{E,F}$ is the normalized Haar measure on $(X, \ast)$. In particular, $\mu_{E,F}$ does not depend on $E, F$. 

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Example 1. Let $X = \Omega_p$, the full shift with $p$ symbols, i.e. $\Omega_p = \{0, 1, ..., p-1\}^{\mathbb{Z}^+}$ with the distance

$$d_q(x, y) = \sum_{i=0}^{\infty} \frac{|x_i - y_i|}{q^i}, \quad q > 1.$$  

$\Omega_p$ can be equipped by the group operation $\oplus$ as follows:

$$(x \oplus y)_i = x_i + y_i, \quad \text{mod } p$$

It is clear that $(\Omega_p, \oplus)$ is a continuous group. Moreover, the right translation by any element is an isometry. Therefore $\mu_{E, F} = \mu$ coincides with the Haar measure which, in fact, is the $(1/p, ..., 1/p)$-Bernoulli measure.

Example 2. Let $X = \Omega_M$ be a topological Markov chain, defined by a finite matrix $M : \{0, 1, ..., p-1\}^2 \to \{0, 1\}$, i.e. $\Omega_M = \{<x_0, x_1, ... | x_i \in \{1, 2, ..., p-1\} \text{ and } M(x_i, x_{i+1}) = 1 \}$ Metric $d$ is the same as in Example 1.

Cylinder $[a_0, a_1, ..., a_{n-1}]$ of the length $n$ is the set of all $x \in \Omega_M$, such that $x_i = a_i$ for $i = 0, 1, ..., n-1$. A word $<a_0, a_1, ..., a_{n-1}>$ is admissible iff $[a_0, a_2, ..., a_{n-1}] \neq \emptyset$. Let $W_n$ be the set of all admissible words of the length $n$ and $\alpha$ be a permutation of $W_n$ such that $(\alpha(w))_{n-1} = w_{n-1}$ for every $w \in W_n$ (admissible permutation). Given such an $\alpha$ define $g_\alpha : X \to X$ as follows:

$$g_\alpha(x) = (a(x_0, x_1, ..., x_{n-1}), x_n, x_{n+1}, ...).$$

It is simple to see that $g_\alpha$ is a local isometry. It implies that $\mu_{E, X}(\{a_0, a_1, ..., a_{n-1}\}) = \mu_{E, X}(\{b_0, b_1, ..., b_{n-1}\})$ if $[a_0, a_1, ..., a_{n-1}] \neq \emptyset$, $[b_0, b_1, ..., b_{n-1}] \neq \emptyset$ and $a_{n-1} = b_{n-1}$. Indeed, under these assumptions there exists an admissible permutation $\alpha : W_n \to W_n$ such that $\alpha(a_0, a_1, ..., a_{n-1}) = b_0, b_1, ..., b_{n-1}$. So, the measure $\mu_{E, X}$ of a nonempty cylinder $\{a_0, a_1, ..., a_{n-1}\}$ depends only on $a_{n-1}$ and $n$. Let

$$v_i(n) = \begin{cases} \mu_{E, X}(\{a_0, a_1, ..., a_{n-2}, i\}) & \text{for an admissible } <a_0, a_1, ..., a_{n-2}, i > \text{ for } i > (v_i(n) = 0 \text{ if there is no admissible words of length } n \text{ ending by } i). \end{cases}$$

It is simple to check that

$$v_i(n) = \sum_{j, M(i, j) = 1} v_j(n + 1)$$

This relation can be rewritten in the matrix form

$$v(n) = Mv(n + 1),$$

where $v(n) = (v_0(n), v_1(n), ..., v_{p-1}(n))^T$ is a column vector. If $M$ is a primitive matrix ($M^p > 0$ for some $p$) then this equation uniquely defines the measure $\mu_{E, X}$, which in this case turns out to be independent of $E, F$. Indeed, by Perron Theorem matrix $M$ has unique positive eigenvector $e$ with eigenvalue $\lambda > 0$ (in our case, in fact, $\lambda > 1$). Let $P$ be the set of all lines in $\mathbb{R}^p$, generated by non-negative vectors. From the proof of Perron Theorem (see, for example, [3])

$$\bigcap_{n \in \mathbb{N}} M^n(P) = \{l_e\},$$

where $l_e$ is a line, generated by $e$. Since $v(n) > 0$ and $v(k) = M^n v(n + k)$, one has $l_e \in M^n(P)$ for any $n$. Hence, $v(k) = \lambda^{-n} e_n$. We have proved the following

Proposition 9 Let $M$ be a primitive matrix and $C \subset \Omega_M$ is an admissible cylinder of length $n$, ending by $a$. Then $\mu_{E, X}(C) = \lambda^{-n} e_n$, where $(e_0, e_1, ..., e_{p-1})$ is the positive eigenvector of $M$, with $e_0 + e_1 + ... + e_{p-1} = 1$.

Example 3. Here we construct an example where $\mu_{E, X}$ is not unique. Let $X = \Omega_{0,1} \cup \Omega_{2,3}$, where $\Omega_{i,j}$ is the Bernoulli shift of symbols $i,j$. We are going to introduce a metric $d$ on $X$ such that $\mu_{E, X}$ depends on $E, F$.

Let us define $d$. For $x \in \Omega_{0,1}$ and $y \in \Omega_{2,3}$ let $d(x, y) = 1$. For $x, y \in \Omega_{0,1}$, $x \neq y$, and $x_i = y_i$, for $i < n$, let $d(x, y) = a_n$. For $x, y \in \Omega_{2,3}$, $x \neq y$, and $x_i = y_i$, for $i < n$, let $d(x, y) = b_n$. Suppose, $1 \geq a_0 \geq a_1 \geq ... \geq a_n \to 0$ and $1 \geq b_0 \geq b_1 \geq ... \geq b_n \to 0$. Straightforward calculations show that $d$ is a metric (even an ultrametric) defining the Markov topology on $X$.

Proposition 10 If $a_{n-1} \geq \epsilon > a_n$ and $b_{n-1} \geq \epsilon > b_n$ then $C_i(\Omega_{0,1}) = 2^\epsilon$ and $C_i(\Omega_{2,3}) = 2^{-\epsilon}$, the cardinality of an optimal $\epsilon$-separated set on $\Omega_{0,1}$ and $\Omega_{2,3}$, correspondingly.
As in the proof of Proposition 6 it is enough to show that for any \( \epsilon \), \( \Omega \), \( \epsilon \), \( n \), \( \epsilon_n = 1/(2n + 1) \). The idea is to choose \( a_n \) and \( b_n \) such that

\[
\frac{C_n^{(2)}(\Omega_{2,3})}{C_n^{(2)}(\Omega_{0,1})} \to 0 \text{ and } \frac{C_n^{(2)}(\Omega_{0,1})}{C_n^{(2)}(\Omega_{2,3})} \to 0,
\]

as \( n \to \infty \). In particular, we can take \( b_0 = 1, a_{(n-1)(2(n-1)+1)} = a_{(n-1)(2(n-1)+1)} = \ldots = a_n/(2n+1) = 1/2n \) and \( b_{n(2n-1)} = b_{n(2n-1)+1} = \ldots = b_{n(n)} = 1/(2n+1) \), where \( n = 1, 2, \ldots \). Now one can check that

\[
a_n(2n+1) - 1 = \frac{1}{2n} = \epsilon_n > a_n(2n+1) \text{ and } b_n(2n+1) - 1 = \frac{1}{2n} = \epsilon_n > b_n(2n+1).
\]

Because of the proposition \( C_n^{(2)}(\Omega_{0,1}) = 2^{n(2n+1)}, C_n^{(2)}(\Omega_{2,3}) = 2^{n(2n+1)} \) and the first limit in (3) occurs.

On the other hand

\[
a_n(2n+1) - 1 > \frac{1}{2n+1} = \epsilon_n' > a_n(2n+1) \text{ and } b_n(2n+1) - 1 = \frac{1}{2n+1} = \epsilon_n' > b_n(2n+1).
\]

So, \( C_n^{(2)}(\Omega_{0,1}) = 2^{n(2n+1)} \) and \( C_n^{(2)}(\Omega_{2,3}) = 2^{n(2n+1)} \); the second limit of (3) is valid. Now, for \( E = \{1/2n \mid n \in \mathbb{N}\} \) one has \( \mu_{E,F}(\Omega_{2,3}) = 0 \) and \( \mu_{E,F}(\Omega_{0,1}) \) is the \((1/2, 1/2)\)-Bernoulli measure, independently of \( F \). For \( E' = \{1/(2n+1) \mid n \in \mathbb{N}\} \) one has \( \mu_{E',F}(\Omega_{0,1}) = 0 \) and \( \mu_{E',F}(\Omega_{2,3}) \) is the \((1/2, 1/2)\)-Bernoulli measure, independently of \( F \). For \( \tilde{E} = \{1/n \mid n \in \mathbb{N}\} = E \cup E' \) the measure \( \mu_{E,F} \) will depend on \( F \).

4 Measures of dual complexity

To define measures of dual complexity we proceed in the same way as in Section 3, just replacing \( \epsilon \)-separated sets by \( \epsilon \)-nets.

Given \( \epsilon > 0 \), consider an optimal \( \epsilon \)-net \( A_{\epsilon} \). Introduce the following functional

\[
\tilde{I}_{\epsilon}(\phi) = \frac{1}{R_{\epsilon}} \sum_{x \in A_{\epsilon}} \phi(x).
\]

Consider

\[
\tilde{I}(\phi) = \lim_{\epsilon \to 0} \tilde{I}_{\epsilon}(\phi).
\]

**Theorem 4** The functional \( \tilde{I} \) is independent of the choice of an optimal \( \epsilon \)-nets \( A_{\epsilon} \).

**Proof.** The proof is similar to the one of Theorem 2 just instead of Proposition 6 one should use Proposition 11 formulated below. \( \square \)

**Proposition 11** Let \( A \) be an optimal \( \epsilon \)-net and \( B \) be an \( \epsilon \)-net. There exists an injective map \( \alpha : A \to B \) such that \( d(x, \alpha(x)) \leq 2\epsilon \) for any \( x \in A \).

**Proof.** Again we will use Marriage Lemma (Lemma 1). For \( x \in A \) let

\[
B_x = \{y \in B \mid O_x(y) \cap O_x(x) \neq \emptyset\} \subseteq O_x(x) \cap B
\]

For \( S \subseteq A \) let

\[
B_S = \bigcup_{x \in S} B_x.
\]

As in the proof of Proposition 6 it is enough to show that for any \( S \subseteq A \) one has

\[
|B_S| \geq |S|.
\]

First of all, \( O_x(x) \subseteq O_x(B_x), x \in A \). Indeed, due to \( O_x(B) = X \) we have

\[
O_x(x) = O_x(x) \cap O_x(B) = O_x(x) \cap O_x(B_x).
\]

So, \( O_x(S) \subseteq O_x(B_S) \).
Now, suppose that \(|B_S| < |S|\) in contradiction to (\(*\)). Then
\[ |A \setminus S \cup B_S| < |A|. \]
Moreover, \(O_i(A \setminus S) \supseteq O_i(A) \setminus O_i(S)\). Indeed, if \(z \in O_i(A)\) and \(z \notin O_i(S)\) then there exists \(a \in A\) such that \(d(a, x) < \epsilon\); \(a\) cannot belong to \(S\) because \(z \notin O_i(S)\). Hence, \(a \in A \setminus S\), and \(z \in O_i(A \setminus S)\). Thus
\[ O_i(A \setminus S \cup B_S) = O_i(A \setminus S) \cup O_i(B_S) \supseteq O_i(A) \setminus O_i(S) \cup O_i(B_S) = X \setminus O_i(S) \cup O_i(B_S) = X, \]
the contradiction with minimality of \(A\).

**Definition 7** The measures \(\nu_{E,F}(X)\) corresponding to \(I_{E,F}\) will be called dual measures of complexity.

**Proposition 12** Let \(\tau\) be an \(\epsilon_0\)-isometry and \(A\) be an \(\epsilon\)-net, \(\epsilon \leq \epsilon_0\). Then \(\tau(A)\) is also an \(\epsilon\)-net.

**Proof.** Given \(x \in X\) we have to prove that \(x \in O_i(\tau(A))\). Due to surjectivity of \(\tau\) there exists \(y \in X\), \(x = \tau(y)\). There exists \(a \in A\) such that \(y \in O_i(a)\). By the definition of \(\epsilon\)-isometry \(x = \tau(y) \in O_i(\tau(a))\). \(\square\)

Using Proposition 12 Proposition 1 one can prove the following analogue of Theorem 5.

**Theorem 5** Let \(\tau\) be a local isometry. Then \(\nu_{E,F}\) is invariant, i.e. \(\nu_{E,F}(A) = \nu_{E,F}(\tau^{-1}(A))\) for all measurable \(A\).

We don’t know if \(\mu_{E,F}\) and \(\nu_{E,F}\) can be different, but we can prove the following theorem.

**Theorem 6** If there exists \(k \in \mathbb{N}\) such that for any \(x \in X\) and any small enough \(\epsilon > 0\) one has \(C_i(O_i(x)) \leq k\), then \(\mu\) and \(\nu\) are equivalent and, moreover,
\[ \frac{1}{k} \nu_{E,F}(A) \leq \mu_{E,F}(A) \leq k \nu_{E,F}(A) \]
for any Borel set \(A \subseteq X\).

It easily implies

**Corollary 2** If \(d\) is a ultrametric (i.e. \(d(x, z) \leq \max\{d(x, y), d(y, z)\}\) for any \(x, y, z \in X\)) then \(\nu_{E,F} = \mu_{E,F}\).

**Proof.** The result follows from the fact that \(C_i(O_i(x)) = 1\) for any \(x \in X\) and any \(\epsilon > 0\), so \(k = 1\) in the conditions of the theorem. Indeed, for any \(z, y \in O_i(x)\) one has \(d(z, y) \leq \max\{d(z, x), d(x, y)\} < \epsilon\) \(\square\)

So, for Example 3 of Section 5 one has \(\nu_{E,F} = \mu_{E,F}\). The measures in Examples 1, 2 are also coincide because of Theorem 5.

In our proof of Theorem 5 we will use the following proposition.

**Proposition 13** Let \(A\) be an optimal \(\epsilon\)-net and \(B\) be an optimal \(\epsilon\)-separated set. Then there exists a collection \(\{K_x\}\) of subsets of \(B\), indexed by elements of \(A\), with the following properties:

- \(K_x \subseteq O_i(x) \cap B\) for any \(x \in A\);
- \(K_x \neq \emptyset\) for any \(x \in A\);
- \(K_x \cap K_y = \emptyset\) for any different \(x, y \in A\);
- \(\bigcup_{x \in A} K_x = B\).

**Proof.** Since \(B\) is an \(\epsilon\)-net, it follows from Proposition 14 that there exists an injective map \(\alpha : A \to B\). So, we can put \(\alpha(x)\) to \(K_x\) and distribute the points \(B \setminus \alpha(A)\) among \(K_x\) so that \(K_x\) satisfy the properties claimed. (For example, we can order \(A\) and put \(b \in B \setminus \alpha(A)\) into \(K_x\) with the smallest \(x \in A\) such that \(b \in O_i(x)\)). \(\square\)

**Proof of Theorem 5** It is enough to show that for non-negative continuous \(\phi\)
\[ \frac{1}{k} \tilde{I}_i(\phi) - \delta_i(\phi) \leq I_i(\phi) \leq k \tilde{I}_i(\phi) + \delta_i(\phi), \]
where \(\delta_i(\phi)\) is the modulus of continuity of \(\phi\). Let \(A\) be an optimal \(\epsilon\)-net and \(B\) be an optimal \(\epsilon\)-separated set. Let \(K_x\) be the sets of Proposition 14. From the conditions of the theorem it follows that \(|K_x| \leq k\) and \(R_x \leq C_i \leq k R_x\). Then
\[ \sum_{y \in B} \phi(y) = \sum_{x \in A} \sum_{y \in K_x} \phi(y) \leq \sum_{x \in A} |K_x| (\phi(x) + \delta_i(\phi)) \leq k \sum_{x \in A} \phi(x) + C_i \delta_i(\phi). \]
Thus
\[
\frac{1}{C_c} \sum_{y \in B} \phi(y) \leq \frac{k}{C_c} \sum_{x \in A} \phi(x) + \delta_c(\phi) \leq \frac{k}{R_c} \sum_{x \in A} \phi(x) + \delta_c(\phi),
\]
that proves the right inequality in (4). Similarly,
\[
\sum_{y \in B} \phi(y) = \sum_{x \in A} \sum_{y \in K_x} \phi(y) \geq \sum_{x \in A} |K_x|(\phi(x) - \delta_c(\phi)) \geq \sum_{x \in A} \phi(x) - C \delta_c(\phi).
\]
Thus
\[
\frac{1}{C_c} \sum_{y \in B} \phi(y) \geq \frac{1}{C_c} \sum_{x \in A} \phi(x) - \delta_c(\phi) \geq \frac{1}{kR_c} \sum_{x \in A} \phi(x) - \delta_c(\phi),
\]
\[\square\]

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