

## SOME CHARACTERISTICS OF THE PROCESS MEASURE OF THE AMOUNT OF INFORMATION

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**Signs and symbols.**  $a = a_1 a_2 \dots a_n$  – binary word of length  $n$ .

$\Lambda$  – empty word.

$X$  – the space of all finite words over  $\{0, 1\}$ . ( $\Lambda \in X$  by definition)

$l(a)$  – the length of word  $a$ .

$\bar{a} = a_1 a_1 a_2 a_2 \dots a_n a_n 01$  – manner of recording the word  $a$  required to record two or more words in the form of one word. For example for the words  $x$ ,  $y$  and  $z$  the record is  $\bar{x}\bar{y}z$ . From the word  $\bar{x}\bar{y}z$  it is possible to decode the words  $x$ ,  $y$  or  $z$  by means of general, recursive functions  $\pi_1$ ,  $\pi_2$  and  $\pi_3$ . (We also have  $\bar{\Lambda} = 01$ .)

$a \subset b$  means  $b = aw$ ,  $w \in X$  ( $aw$  is a concatenation of words  $a$  and  $w$ ).

$f(x) \preccurlyeq g(x)$  means  $(\exists C)(\forall x \in X) f(x) \leq g(x) + C$ .

$f(x) \asymp g(x)$  means  $f(x) \preccurlyeq g(x)$  and  $g(x) \preccurlyeq f(x)$ .

The function  $F(a_1 a_2 \dots a_n) = 2^n - 1 + \sum_{i=1}^n a_i 2^{n-i}$  gives a one-to-one correspondence of the set  $X$  and the set  $\{0, 1, 2, \dots\}$ . The symbol  $a$  will denote both the word and its corresponding number.

**Introduction.** The partial recursive function  $\mathcal{F} : X^{m+1} \rightarrow X$  of  $m + 1$  arguments is called a process according to argument  $p$  if the following applies: for a word  $p$ ,  $\mathcal{F}(p, y_1, \dots, y_m)$  exists and if  $q \subset p$ , then  $\mathcal{F}(q, y_1, \dots, y_m)$  exists and  $\mathcal{F}(q, y_1, \dots, y_m) \subset \mathcal{F}(p, y_1, \dots, y_m)$ .

*Definition 1.* The conditional process complexity of  $(x_1, \dots, x_n)$ , given  $(y_1, \dots, y_m)$ , with respect to the processes  $\mathcal{F}_1, \dots, \mathcal{F}_n$  is

$$\begin{aligned} KP_{\mathcal{F}_1, \dots, \mathcal{F}_n}(x_1, \dots, x_n / y_1, \dots, y_m) &= \\ &= \min_{p \in X} \{ \alpha(p) / \mathcal{F}_1(p, y_1, \dots, y_m) = x_1, \dots, \mathcal{F}_n(p, y_1, \dots, y_m) = x_n \}. \end{aligned}$$

The function  $\alpha(p)$  is a criterion of complexity and it is usually taken as  $\log_2 p$ , which in the alphabet  $0 - 1$  is equal to  $l(p) + C$ .

**THEOREM 1.** *There is a set of optimal  $m + 1$  dimensional processes according to argument  $p(\mathcal{F}^\circ(p, y_1, \dots, y_m), \dots, \mathcal{F}_n^\circ(p, y_1, \dots, y_m))$  such that for any other set of  $m + 1$  dimensional processes according to argument  $p(\mathcal{G}_1(p, y_1, \dots, y_m), \dots, \mathcal{G}_n(p, y_1, \dots, y_m))$  and for any  $(x_1, \dots, x_n)$*

$$KP_{\mathcal{F}_1^\circ, \dots, \mathcal{F}_n^\circ}(x_1, \dots, x_n/y_1, \dots, y_m) \preceq KP_{\mathcal{G}_1^\circ, \dots, \mathcal{G}_n^\circ}(x_1, \dots, x_n/y_1, \dots, y_m).$$

The proof of Theorem 1. is standard for this theory and similar with the proof in [2, p. 91 Theorem 1.2].

From now on, the complexity  $KP_{\mathcal{F}_1^\circ, \dots, \mathcal{F}_n^\circ}(x_1, \dots, x_n/y_1, \dots, y_m)$  will be designated with  $KP(x_1, \dots, x_n/y_1, \dots, y_m)$ .  $KP(x_1, \dots, x_n)$  means  $KP(x_1, \dots, x_n/\Lambda, \dots, \Lambda)$ .

We have the following characteristics of the process complexity:

$$(i) \quad KP(x/y) \preceq KP(x) \preceq KP(x/y) + 2KP(y)$$

where  $K(y)$  is the Kolmogorov complexity of the word  $y$ . Let  $KP(x/y) = l(p)$ , that is,  $\mathcal{F}^\circ(p, y) = x$ . Let us form the function

$$S = \begin{cases} \mathcal{F}^\circ(\pi_2(z), F^\circ(\pi_1(z))), & \text{if } z \text{ has the form } \bar{a}b \\ \Lambda, & \text{otherwise.} \end{cases}$$

$\mathcal{F}^\circ$  is an optimal two-dimensional process, and  $F^\circ$  an optimal function for Kolmogorov complexity. Let  $K(y) = l(p_y)$ . The function  $S$  is a process by construction. For the program  $z = \bar{p}_y p$  the results is  $x$ . Further more, we have

$$KP(x) \preceq KP_{\mathcal{G}}(x) \leq l(\bar{p}_y) + KP(x/y) \asymp KP(x/y) + 2K(y).$$

*Remark.* The constant 2 may be replaced with  $1 + \varepsilon$  by a more appropriate coding of the program  $z$ .

$$(ii) \quad KP(x/y) \preceq K(x/y) + 2\log_2 K(x/y)$$

Let us form a process

$$\mathcal{J}^2(z, y) = \begin{cases} F^\circ(A(z), y), & \text{if } z \text{ has the form } \bar{a}b \text{ and } l(b) \geq a \\ \Lambda, & \text{otherwise} \end{cases}$$

where  $A(\overline{l(p)}pq) = p$  is general recursive ( $p, q \in X$ ). For  $F^\circ(p_x, y) = x$  and  $z = \overline{l(p_x)}px$  we have

$$KP(x/y) \preceq KP_{\mathcal{J}}(x/y) \leq l(z) = l(\overline{l(p_x)}) + K(x/y) \asymp K(x/y) + 2l(K(x/y)).$$

$$(iii) \text{ If } \mathcal{F}(x) \text{ is a process, then } KP(\mathcal{F}(x)) \preceq KP(x).$$

If for  $\mathcal{F}(x)$  there exists an inverse function that is also a process, then  $KP(\mathcal{F}(x)) \asymp KP(x)$ .

$$(iv) \quad KP(x/y) \succcurlyeq KP(x/y, z) \tag{1.1}$$

$$\begin{aligned} KP(x/y) &= \min\{l(p)/\mathcal{F}^\circ(p, y) = x\} = \min\{l(p)/\mathcal{G}(p, y, z) = x\} \succcurlyeq \\ &\succcurlyeq \min\{l(p)/\mathcal{F}^\circ(p, y, z) = x\} = KP(x/y, z). \end{aligned}$$

The function  $\mathcal{G}(p, y, z) = \mathcal{F}^\circ(p, y)$  has  $z$  as a fictive argument.

(v) For every partial recursive function  $F$  we have

$$\begin{aligned} KP(y/x, F(x)) &\asymp KP(y/x) \\ KP(y/x, F(x)) &= \min\{l(p)/\mathcal{F}^\circ(p, x, F(x)) = y\} = \\ &= \min\{l(p)\mathcal{G}(p, x) = y\} \asymp KP(y/x). \end{aligned}$$

(vi) If  $F$  is an invertible partial recursive function, then

$$KP(x/F(x)) \asymp KP(F(x)/x) \asymp 0 \tag{1.2}$$

$$KP(x/F(x)) = \min\{l(p)/\mathcal{F}^\circ(p, F(x)) = x\} \preccurlyeq \min\{l(p)/\mathcal{G}(p, F(x)) = x\} \asymp 0,$$

where  $\mathcal{G}(p, F(x)) = F^{-1}(F(x))$ , which is trivially a process according to  $p$ .

$$KP(F(x)/x) = \min\{l(p)/\mathcal{F}^\circ(p, x) = F(x)\} \preccurlyeq \min\{l(p)/\mathcal{G}(p, x) = F(x)\} \asymp 0,$$

where  $\mathcal{G}(p, x) = F(x)$ , which is also a process according to  $p$ .

**Measure of the amount of information.** The process complexity of a word  $x$  is very suitable for defining the concept of randomness. Namely, (Schnorr in [4] shows that to a Martin-Löf random binary sequences  $\omega$  applies  $KP(\omega^n) \asymp n$ , where  $\omega^n$ , is a fragment of the sequences  $\omega$  of length  $n$ . On the other hand, the complexity is also suitable for defining the measure of information. Kolmogorov defines in [1] the measure of information carried by a word  $y$  about word  $x$  as

$$I(y : x) = K(x) - K(x/y) \tag{2.1}$$

Levin ([5]) also defines the measure of information as  $IP(y : x) = \overline{KP}(x) - \overline{KP}(x/y)$ , where  $\overline{KP}_A(x) = \min\{l(p)/A(p) = x\}$  and  $A(p)$  is a function such if  $A(p) = x$ , then  $A(pq) = x$ . (Those are the so-called prefix algorithms.)

*Definition 2.* The quantity

$$\begin{aligned} J(y_1, \dots, y_m : x_1, \dots, x_n / z_1, \dots, z_k) &= KP(x_1, \dots, x_n / z_1, \dots, z_k) - \\ &- KP(x_1, \dots, x_n / y_1, \dots, y_m, z_1, \dots, z_k) \end{aligned}$$

is termed the process measure of the amount of information that  $(y_1, \dots, y_m)$  carries on  $(x_1, \dots, x_n)$  if  $(z_1, \dots, z_k)$  is known. We have the following characteristics of measure  $J$ :

$$(i) \quad J(y : x) \asymp 0 \tag{2.2}$$

The property (2.2) follows from the relation (1.1).

$$(ii) \quad J(x : x) \asymp KP(x) \tag{2.3}$$

The relation (2.3) is a direct consequence of (1.15). It can be also shown that  $J(p_x : x) \asymp KP(x)$ , where  $p_x$  is such that  $\mathcal{F}^\circ(p_x) = x$ .

$$(iii) \quad J(x, y : z) = J(x : z) + J(y : z/x) \tag{2.4}$$

The proof results directly from the definition of the measure  $J$ .

(iv) The process measure of information may be compared with measure  $I$ , introduced by (2.1)

$$\begin{aligned} I(y : x) - 2 \log_2 K(x/y) &\preceq J(y : x) \preceq I(y : x) + 2 \log_2 K(x) \\ J(y : x) = KP(x) - KP(x/y) &\preceq K(x) + 2 \log_2 K(x) - K(x/y) = \\ &I(y : x) + 2 \log_2 K(x). \end{aligned}$$

(v) If  $F$  is partial recursive and invertible function,  $J(F(x) : x) \asymp KP(x)$ ,  $J(x : F(x)) \asymp (F(x))$ ,  $J(F(x) : y) \asymp J(x : y)$ .

(vi) It is known that the algorithm measure of the amount of information is not commutative ([2], [3]), that is, it can be shown only as  $|J(y : x) - I(x : y)| \preceq 12 \cdot I(K(x, y))$ . Since  $|J(y : x) - I(y : x)| \preceq (1 + \varepsilon)l(K(x))$ , for the process measure  $J$  we have

$$|J(y : x) - J(x : y)| \preceq (14 + 2\varepsilon)l(K(x, y)).$$

(vii) For every word  $x$  we have  $J(l(x) : x) \preceq 2 \cdot K(l(x))$ .

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