

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 .

Λ – 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 π_1, π_2 and π_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).

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

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

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

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

\mathcal{F}° is an optimal two-dimensional process, and F° an optimal function for Kolmogorov complexity. Let $K(y) = l(p_y)$. The function \mathcal{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 $A(\bar{l}(p)pq) = p$ is general recursive ($p, q \in X$). For $F^\circ(p_x, y) = x$ and $z = \bar{l}(p_x)p_x$ we have

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

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

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

$$KP(x/F(x)) \asymp KP(F(x)/x) \asymp 0 \quad (1.2)$$

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

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

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

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 ω applies $KP(\omega^n) \asymp n$, where ω^n , is a fragment of the sequences ω 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

$$I(y : x) = K(x) - K(x/y) \quad (2.1)$$

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 :

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

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

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

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 $\mathcal{F}^\circ(p_x) = x$.

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

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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