On Kolmogorov Structure Functions

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Abstract

All strings with low mutual information with the halting sequence will have flat Kolmogorov Structure Functions, in the context of Algorithmic Statistics. Assuming the Independence Postulate, strings with non-negligible information with the halting sequence are purely mathematical constructions, and cannot be found in the physical world.

1 Introduction

In statistics, one tries to determine a model (such as a parameter for a distribution) from data which is assumed to have noise. In the Minimum Description Principle [Gru07], the model that describes information with the shortest code is assumed to be the best model. The data is described as a two part code, where the first part is the model and the second part is the noise. In one of his last works, Kolmogorov suggested a two part code for individual strings $x \in \{0,1\}^*$ based off Kolmogorov Complexity. The first part (the model) is a set D containing x, the second part (the noise) is the code of x given D, of size $\lceil \log |D| \rceil$. Other works examined probabilities and also total computable functions as models $\lceil \text{Vit02} \rceil$. Kolmogorov suggested the following structure function at the Tallinn conference in Estonia, 1973.

$$\mathbf{H}_k(x) = \{ \log |S| : x \in S, \mathbf{K}(S) \le k \}.$$

The function **K** is the prefix Kolmogorov complexity. This definition is used for the following function, which is a central definition of *Algorithmic Statistics* [VS15, VS17, VV04],

$$k \mapsto k + \mathbf{H}_k(x) - \mathbf{K}(x).$$

This function's equivalence to several other definitions is the main theorem of Algorithmic Statistics [SSV24]. Furthermore, Theorem 1 of [VS17] showed that any shape of the structure function is possible.

The structure function is flat for all strings with low mutual information with the halting sequence. Assuming the *Independence Postulate*, [Lev84, Lev13], strings with non-negligible mutual information with the halting sequence are exotic, in that they cannot be found in nature. Such strings are purely mathematical constructions.

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2 Bounds

We review the results of [GTV01], in particular Theorem of III.24, which I don't think is widely known in the literature. $\mathbf{m}(x)$ is the algorithmic probability. The amount of information that the halting sequence $\mathcal{H} \in \{0,1\}^{\infty}$ has about $x \in \{0,1\}^*$ is $\mathbf{I}(x;\mathcal{H}) = \mathbf{K}(x) - \mathbf{K}(x|\mathcal{H})$. We use $x <^+ y, x >^+ y$ and $x =^+ y$ to denote x < y + O(1), x + O(1) > y and $x = y \pm O(1)$, respectively. In addition, $x <^{\log} y$ and $x >^{\log} y$ denote $x < y + O(\log y)$ and $x + O(\log x) > y$, respectively. For $x, y \in \{0,1\}^*$, $x \sqsubseteq y$ if y = xz for some $z \in \{0,1\}^*$. [A] = 1 if mathematical statement A is true, and [A] = 0 otherwise.

Let $S_k = \{x : \mathbf{K}(x) \leq k\}$. Let $N_k = |S_k|$ where $\log N_k = {}^+k - \mathbf{K}(k)$, due to [GTV01]. Let I_k^x be the index of x in an enumeration of S_k . For $\mathbf{K}(x) = k$, let m_x be the longest joint prefix of I_k^x and N_k . So $m_x 0 \sqsubseteq I_k^x$ and $m_x 1 \sqsubseteq N_k$. Let $S_x = \{y : m_x 0 \sqsubseteq I_k^y\}$. So

$$\log |S_x| =^+ k - \mathbf{K}(k) - ||m_x||$$

$$\mathbf{K}(S_x) <^+ \mathbf{K}(k) + \mathbf{K}(m_x) <^+ \mathbf{K}(k) + ||m_x|| + \mathbf{K}(||m_x||).$$

Theorem 1 ([GTV01]).

$$||m_x|| < \mathbf{K}(\mathbf{K}(x)) + \mathbf{I}(x; \mathcal{H}) + O(\log \mathbf{I}(x; \mathcal{H})).$$

Proof. Let $\nu(y) = c[\mathbf{K}(y) \leq k]\mathbf{m}(y)2^{\|m_y\|}/(\|m_y\|^2)$. For proper choice of c, ν is a semimeasure and computable relative to \mathcal{H} and k. So $\mathbf{K}(x|\mathcal{H},k) <^+$ $-\log \nu(x) =^+ \mathbf{K}(x) - \|m_x\| + 2\log \|m_x\|$.

Corollary 1. For $x \in \{0,1\}^*$, $n = \mathbf{K}(x)$, for all $m \le n$, $m \in \mathbb{W}$, there is a set $S \ni x$ such that $|S| = 2^m$ and $\mathbf{K}(S) + m <^{\log} n + \mathbf{I}(x; \mathcal{H})$.

Claim 1. Thus there exists a set $S \ni x$ such that $\mathbf{K}(S) < 2\mathbf{K}(\mathbf{K}(x)) + \mathbf{I}(x; \mathcal{H}) + O(\log \mathbf{I}(x, \mathcal{H}))$ and $\mathbf{K}(S) + \log |S| <^+ \mathbf{K}(x) + \mathbf{K}(\mathbf{K}(x)) + O(\log \mathbf{I}(x; \mathcal{H}))$. This fact combined with the following proposition characterizes the Kolmogorov Structure Function.

Proposition 1. Let $S \ni x$. For all $s < \log |S|$ there exists a set $S' \ni x$ such that $|S'| \le |S|2^{-s}$ and $\mathbf{K}(S') <^+ \mathbf{K}(S) + s + \mathbf{K}(s)$.

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