

# On one-way functions and the average time complexity of almost-optimal compression

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#### Abstract

We show that one-way functions exist if and only there exists an efficient distribution relative to which almost-optimal compression is hard on average. The result is obtained by combining a theorem of Ilango, Ren, and Santhanam [\[IRS21,](#page--1-0) [IRS22\]](#page--1-1) and one by Bauwens and Zimand [\[BZ23\]](#page--1-2).

### 1 Introduction

Several recent papers show that the existence of one-way functions (OWF) is equivalent to the hardness of certain problems in meta-complexity [\[LP20,](#page--1-3) [LP21,](#page--1-4) [RS21,](#page--1-5) [IRS21,](#page--1-0) [IRS22,](#page--1-1) [LP23a,](#page--1-6) [LP23b,](#page--1-7) [HLO24,](#page--1-8) LS24. The motivation for this research line comes primarily from cryptography, where one-way func-tions play a central role<sup>[1](#page-0-0)</sup>. Ilango, Ren and Santhanam [\[IRS21,](#page--1-0) [IRS22\]](#page--1-1) have obtained a result of this type involving standard (unbounded) Kolmogorov complexity. Informally speaking, they have shown that one-way functions exist if and only if "finding good approximations of Kolmogorov complexity" is hard on average with respect to some polynomial-time samplable distribution. Bauwens and Zimand [\[BZ23\]](#page--1-2) have shown that given a good approximation of the Kolmogorov complexity of a string  $x$ , one can compress  $x$  in probabilistic polynomial time to a string of length close to its complexity (so,  $x$  is almost-optimally compressed). The combination of these 2 results yields the following theorem.

Theorem 1 (Informal statement). The following two assertions are equivalent:

- 1. There exists an one-way function.
- 2. Almost optimal compression is hard on average with respect to some polynomial-time samplable distribution.

The result of Ilango et. al. [\[IRS21,](#page--1-0) [IRS22\]](#page--1-1) is not exactly stated in the form that we mentioned above. For this reason, we prefer to give a proof which does not directly invoke [\[IRS21,](#page--1-0) [IRS22\]](#page--1-1), but which closely follows their method. In one direction, it is based on results of Impagliazzo, Levin and Luby[\[IL90,](#page--1-10) [IL89\]](#page--1-11) connecting the existence of OWFs to the hardness of approximating poly-time samplable distributions, and, in the other direction, it is based on the connection between OWFs and pseudo-random generators established by Håstad, Impagliazzo, Levin and Luby [\[HILL99\]](#page--1-12).

## 2 Definitions, and technical tools

Kolmogorov complexity. We fix an optimal universal Turing machine U with prefix-free domain. A program for string x is a string p such that  $U(p) = x$ . The prefix-free Kolmogorov complexity  $K(x)$ of the string x is the length of a shortest program for  $x^2$  $x^2$ .

**Distributions.** We consider ensembles of distributions. An ensemble has the form  $D = (D_n)_{n \in \mathbb{N}}$ , where each  $D_n$  is a distribution on  $\{0,1\}^n$ . The ensemble D is *samplable* if there exists a probabilistic algorithm Samp, such that for every n and every  $x \in \{0,1\}^n$ ,

$$
Prob [Samp(1^n) = x] = D_n(x)
$$

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<span id="page-0-1"></span><span id="page-0-0"></span><sup>&</sup>lt;sup>1</sup>See [\[HLO24\]](#page--1-8) and [\[LS24\]](#page--1-9) for a discussion of some of these and related works.

<sup>&</sup>lt;sup>2</sup>The prefix-free Kolmogorov complexity  $K(x)$  is a little more convenient for the proof than the plain complexity  $C(x)$ . The difference  $K(x) - C(x)$  is bounded by  $2 \log |x|$  and, therefore, the result is valid for  $C(x)$  as well.

(the probability is over the randomness of Samp).

D is said to be *P-samplable*, in case **Samp** runs in polynomial time.

Some notation: For every x, we denote  $D(x) = D_{|x|}(x)$ . For every m,  $U_m$  denotes the uniform distribution over m-bit strings.

<span id="page-1-1"></span>**Lemma 1.** If  $D$  is samplable, then for every  $x$  in its support,

$$
K(x) \le \log \frac{1}{D(x)} + 3\log(|x|) + O(1).
$$

*Proof.* Fix a binary string x and let n be its length. Given n and the code of  $Samp$ , one can compute  $D_n(y)$  for all strings y of length n and then list all these strings in descending order of their  $D_n(\cdot)$ probability (with ties broken, say, lexicographically). The string x is described by its rank t in this list. Since the  $D_n$ -probability of the first t strings in the order is at most 1 and at least  $t \cdot D_n(x)$ , it follows that  $t \leq [1/D_n(x)]$ . An overhead of  $2\log(|x|) + O(1)$  bits is added to obtain a self-delimited description in the standard way.  $\Box$ 

<span id="page-1-0"></span>**Lemma 2.** For every distribution D, and every  $\Delta \geq 0$ ,

$$
\text{Prob}_{x \leftarrow D} \left[ K(x) \ge \log \frac{1}{D(x)} - \Delta \right] \ge 1 - 2^{-\Delta}.
$$

*Proof.* The complement of the event in the probability is  $E = \{x \mid D(x) \leq 2^{-\Delta} \cdot 2^{-K(x)}\}\.$  We have

$$
D(E) = \sum_{x \in E} D(x) \le \sum_{x \in E} 2^{-\Delta} \cdot 2^{-K(x)} \le 2^{-\Delta} \sum_{x \in \{0,1\}^*} 2^{-K(x)} \le 2^{-\Delta} \cdot 1 = 2^{-\Delta}.
$$

In the penultimate transition, we have used the Kraft inequality, which is legitimate because  $K(\cdot)$ represents the lengths of a prefix-free code.  $\Box$ 

Formal statement of Theorem [1.](#page--1-13) The following 2 assertions are equivalent:

Assertion (1): The hypothesis "  $\exists$  OWF": There exists a polynomial-time computable f:  ${0,1}^* \rightarrow {0,1}^*$  with the following property: For every probabilistic polynomial-time algorithm Inverter, every  $q \in \mathbb{N}$  and every length  $n \in \mathbb{N}$ ,

$$
\text{Prob}_{x \leftarrow U_n, \text{Inverter}}[\text{Inverter}(1^n, f(x)) \in f^{-1}(f(x))] \le 1/n^q.
$$

(The notation Prob<sub>x←Un</sub>,Inverter means that the probability is over  $U_n \times$  randomness of Inverter.)

Assertion  $(2)$ : The hypothesis "almost optimal compression is hard on average" : There exists a P-samplable distribution  $D$  and a constant c with the following property: For every probabilistic polynomial-time algorithm Compress, at every length  $n$ ,

 $\text{Prob}_{x \leftarrow D_n, \text{Compress}}[\textsf{Compress}(x) \text{ outputs a program of } x \text{ of length } \leq K(x) + c \log^2 n] \leq 1/n.$ 

Remark. The "infinitely often" version of Theorem [1](#page--1-13) is also true, with essentially the same proof. More precisely, if we modify Assertions 1 and 2 by replacing "every length  $n$ " with "infinitely many lengths n," the modified assertions are also equivalent.

#### <span id="page-1-2"></span>Results from the literature that we use.

Theorem 2 ( $[IL89, IL90]$  $[IL89, IL90]$  $[IL89, IL90]$ ; this variant is stated and proved in  $[IRS21]$ ). Assume the hypothesis "∃ OWF" is not true. Let  $D = (D_n)_{n \in \mathbb{N}}$  be a P-samplable ensemble of distributions, and  $q \in \mathbb{N}$ . There exists a probabilistic polynomial-time algorithm A and a constant  $c > 1$  such that for infinitely many  $\overline{n},$ 

$$
\text{Prob}_{x \leftarrow D_n, A} [D_n(x)/c \le A(x) \le D_n(x)] \ge 1 - \frac{1}{n^q}.
$$

<span id="page-1-3"></span>In other words: If there are no one-way functions, then P-samplable distributions can be approximated efficiently in the average sense.

**Theorem 3** ([\[BZ23\]](#page-3-3)). There exists a probabilistic polynomial-time algorithm Compress that for every input triple  $(x \in \{0,1\}^*, m \in \mathbb{N}$ , rational  $\epsilon > 0$ ) outputs with probability 1 a string z of length  $m +$  $O(\log m \cdot \log |x|/\epsilon)$  and if  $m \geq K(x)$  then

Prob[z is a program for  $x$ ]  $\geq 1 - \epsilon$ .

In other words: Given a good approximation of the Kolmogorov complexity of a string  $x$ , one can efficiently compress x almost optimally (where efficiently means in probabilistic polynomial time).

## 3 Proof of Theorem [1](#page--1-13)

#### Proof of assertion  $(2) \rightarrow$  assertion  $(1)$ .

We actually prove the contrapositive:  $\sharp$  OWF  $\Rightarrow$  almost optimal compression is easy on average.

Let  $D = (D_n)_{n \in \mathbb{N}}$  be a P-samplable ensemble and  $q \in \mathbb{N}$ . By Lemma [2](#page-1-0) and Lemma [1,](#page-1-1) for some constant  $c$ , for every  $n$ 

$$
\text{Prob}_{x \leftarrow D_n} [\log \frac{1}{D_n(x)} - c \log n \le K(x) \le \log \frac{1}{D_n(x)} + c \log n] \ge 1 - 1/n^q.
$$

Under our assumption " $\exists$  OWF," Theorem [2](#page-1-2) states that there exists an algorithm A that approximates  $K(x)$ . By rescaling we get that, for every n,

$$
Prob_{x \leftarrow D_n}[K(x) \le A(x) \le K(x) + c \log n] \ge 1 - 1/n^q.
$$

Using algorithm Compress from Theorem [3](#page-1-3) with  $m = A(x)$  and  $\epsilon = 1/n^{2q}$ , and after more rescaling we obtain the converse of assertion (2).

#### Proof of assertion  $(1) \rightarrow$  assertion  $(2)$ .

 $(\exists \text{OWF} \Rightarrow \text{almost optimal compression is hard on average})$ 

The idea is that an efficient good compressor would break the security of any candidate pseudorandom generator (p.r.g.), because the output of the generator can be compressed to a much shorter string, whereas a genuinely random string cannot. Therefore, pseudorandom generators would not exist and hence there would be no OWF, contradicting assertion (1). Now, the details.

Suppose " $\exists$  OWF" is true. Then, by [\[HILL99\]](#page-3-4) combined with the methods to obtain ensembles of p.r.g.'s with every possible output length [\[Gol01,](#page-3-5) Section 3.3.3], there exists an ensemble of p.r.g.'s  $G = (G_n)_{n \in \mathbb{N}}$ , with  $G_n : \{0,1\}^{s(n)} \to \{0,1\}^n$ , where the seed length  $s(n)$  is bounded by  $n^{1/3}$ , that satisfies the following security guarantee: For every probabilistic polynomial-time algorithm T (the hypothetical distinguisher) and every  $q \in \mathbb{N}$ , for every  $n \in \mathbb{N}$ , the probabilities that (a) T accepts  $G_n(U_{s(n)})$  and (b) T accepts  $U_n$ , differ by at most  $1/n^q$ .

Consider the following P-samplable distribution  $D_n$ :

with probability 1/2, output  $G(U_{s(n)})$  and with probability 1/2, output  $U_n$ .

Clearly, if assertion  $(2)$  is false, then there exists a probabilistic polynomial-time algorithm A that, at infinitely many lengths n, with  $D_n$ -probability  $\geq 1/n$  approximates  $K(x)$  with slack at most  $c \log^2 n$ . Then, for infinitely many  $n$ , by Markov's inequality,

$$
Prob_{x \leftarrow D_n}[\mathcal{E}_n] \ge 5/6,
$$

where

$$
\mathcal{E}_n = \{x \in \{0,1\}^n \mid \text{Prob}[|A(x) - K(x)| \le c \log^2 n] \ge 6/(5n)\}.
$$

For infinitely many n, the complement of  $\mathcal{E}_n$  (an event that we henceforth denote by BAD and which says that A fails to approximate K), has  $D_n$ -probability at most 1/6. By inspecting the sampling procedure, we see that each element x in BAD has  $D_n$ -probability mass at least  $(1/2) \cdot 2^{-n}$  and thus  $1/6 \ge D_n(BAD) \ge (\#BAD) \cdot (1/2 \cdot 2^{-n}) = 1/2 \cdot Prob_{U_n}[BAD],$  and so

$$
Prob_{U_n}[BAD] \leq 1/3.
$$

Also, each element in BAD ∩ Im( $G(U_{s(n)})$  has  $D_n$ -probability mass at least  $(1/2) \cdot 2^{-s(n)}$ , which, similarly to the above, implies that

$$
Prob_{U_{s(n)}}[BAD \cap Im(G(U_{s(n)}))] \leq 1/3.
$$

We now define the probabilistic polynomial-time distinguisher  $T: T$  on input z of length n outputs 1, if  $A(z) \leq 2s(n)$ , and 0 otherwise. Note that  $G(U_{s(n)})$  with probability 1 has prefix-free complexity at most  $s(n)+2\log s(n)+O(1) \leq n^{1/3}+O(\log n)$ , and  $U_n$ , with probability at least 1-1/n, has complexity at least  $n - \log n$ .

Therefore, for infinitely many  $n$ ,

$$
Prob_{U_{s(n)},T}[T(G(U_{s(n)})) = 1] \ge (1 - 1/3) \cdot 6/(5n)
$$

(the probability that  $G(U_{s(n)})$  is not in BAD is at least  $1 - 1/3$  and the probability that T uses good randomness is at least  $6/(5n)$  and, by similar arguments,

$$
Prob_{U_n,T}[T(U_{4s(n)})=1] \le (1/n+1/3) \cdot 6/(5n),
$$

contradicting the security of G.

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